



Face Recognition under Different Lighting Conditions Using Multiple Feature Fusion

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ABSTRACT: Face recognition has received substantial attention from researchers in biometric pattern recognition. The common challenges faced by the face recognition technology are the range of appearance variations due to changes in illumination, pose, facial expression, aging etc. The objective of this paper is to make face recognition reliable under uncontrolled lighting conditions. This is achieved by an integrated framework consisting of illumination normalization, local texture based face representation, robust feature extraction and multiple feature fusion. Each stage increases resistance to illumination variations. Local ternary pattern, local binary pattern and gabor wavelets are used to extract local feature sets. In the proposed method, an additional feature set is extracted using Elongated Multi Block Local Ternary Pattern (EMB-LTP). It improves the reliability of the algorithm. This multiple feature fusion technique can achieve a higher acceptance ratio and a lower rejection ratio compared to the cases where a single feature set is used.

KEYWORDS: Biometric pattern recognition; Local Binary Pattern; Local Ternary Pattern, Gabor wavelets

I. INTRODUCTION

Face Recognition has gained a substantial place in the area of research in computer vision and image processing during the past several decades. It is one of the most important tasks, and as such has received a great deal of attention because of its applications in various domains such as automated crowd surveillance, access control, mug shot identification, face reconstruction, design of human computer interface and multimedia communication[1].

In general, there are several challenges that one can encounter during the process of facial recognition. They are caused due to variation in illumination, pose, age, facial expressions, aging and partial occlusions. This paper makes an attempt to overcome the challenge of correctly recognizing a face when presented in different illumination conditions. Such variations are overcome by the proposed method involve facial features extracted from multiple methods.

Computer recognition of face images involves two crucial aspects: facial feature representation and classifier design [3]. Facial representation is to derive a set of features from the original face for describing faces. If inadequate features used, even the best classifier will fail to achieve facial recognition. Therefore it is essential to extract discriminative facial features for facial representation. A set of good facial features are said to consist of following properties: first, can tolerate within class variations whilst discriminate different classes well; second, can easily be extracted from the raw images to allow fast processing; and finally, lie in a space with the low dimensionality in order to avoid computationally expensive classifiers.

A single feature extraction technique generally cannot gather both the macro and micro structural information associated with the image. Hence the concept of multiple feature fusion comes into play to create a set of facial features that overcome the former complaints. In this paper, few feature extraction techniques are utilized to get a fused outcome that can assist in facial recognition in a more outstanding way than when the techniques are implemented individually. Those techniques comprise of Local Binary Pattern, Local Ternary Pattern, Gabor filter, Elongated Multi Block Local Ternary Pattern. The classifier used in our paper is PCA[6].

The remainder of this paper is organized as follows. Section II reviews about the various facial feature extraction methods used. Section III explains the proposed method of facial recognition. Section IV reports the results of our experiments on extended Yale B database and discussion of these results. Section V gives the conclusion of this paper.

International Journal of Innovative Research in Computer and Communication Engineering

(An ISO 3297: 2007 Certified Organization)

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Vol. 5, Issue 8, August 2017

II. RELATED WORK

Of the mentioned facial feature extraction methods, Local Binary Pattern (LBP) is the oldest and was introduced by Ojala for the aim of texture classification. It has been extensively exploited in many applications. For instance, face image analysis, image and video retrieval, environment modeling, visual inspection, motion analysis, biomedical and aerial image analysis, and remote sensing. The approach followed is that proposed by Ahonen et al. The face image is divided into a grid of small of nonoverlapping regions, where a histogram of the LBP code for each region is constructed. The similarity of two images is then computed by summing the similarity of histograms from corresponding regions.

LBP method is not only relatively simple and with low computation complexity, but also has the properties of rotational invariance, gray scale invariance and others significant advantages. Despite the great success of LBP in early applications, its practical results are not satisfactory in different fields. Hence, many researchers have improved the LBP and achieved lots of significant results. Some of these variants are: The elongated LBP (ELBP) method for face recognition proposed by Liao et al. The local ternary patterns (LTP) introduced by Tan et al. which is less sensitive to noise in uniform regions. LTP has stronger discrimination ability against the changes of noise and illumination than LBP in the uniform region. But the issues of image multi-scale variation and partial occlusion are needed to solve.

The EMB-LTP (Elongated Multi Block Local Ternary Pattern) is a novel variant of LBP capturing the prominent features of ELBP, MB-LBP and LTP which themselves are variants of LBP developed at different stages. The ELBP features are more generic in nature than the conventional LBP. It is able to capture anisotropic information from the facial images, which are important features as there are many important parts of the face such as eyes, mouth are all elongated structures. The multi-scale block local binary pattern (MB-LBP) encodes not only microstructures but also macrostructures of image patterns. The Elongated LTP (ELTP), another variation of the basic LTP is performed best than basic LBP. Another variation of LBP called multi-scale block local ternary pattern proposed by Xiaofei Jia et al. for fingerprint vitality detection and results show its superiority than the basic LBP and by Zhe Wei et al. for face detection.

Principal Component Analysis (PCA) is an unsupervised dimensionality reduction technique that is used for compression and recognition problems. It is also known as Eigenspace projection or Karhunen-Loeve Transformation. It is used to find a subspace whose basis vectors correspond to maximum variance directions in original space.

III. PROPOSED ALGORITHM

In this paper, the algorithm used is mainly divided into three stages such as Illumination normalization, robust feature extraction followed by subspace representation.

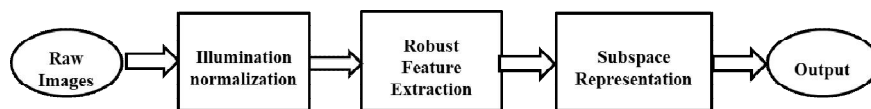


Fig1: Stages of our face recognition method

3.1 Illumination Normalisation

The source of variation in facial appearance is the illumination of the face. The majority of the techniques proposed to deal with this problem exploit the low dimensionality of the face space under varying illumination conditions. They either use several images of the same person recorded under varying illumination conditions or it relies on the availability of 3D face models and different maps to generate novel views. The main shortcoming of this approach is the requirement in practice of large example sets to achieve good reconstructions. Our approach on the other hand

International Journal of Innovative Research in Computer and Communication Engineering

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builds an illumination varying subspace by constructing artificially illuminated colour images from an original image. The robustness of lighting conditions is achieved using pre-processing chain.

3.1.1 Pre-processing chain

Pre-processing is an operation with images at the lowest level of abstraction, both input and output are intensity images. Image pre-processing methods generally involve the use of considerable redundancy in images. Neighboring pixels corresponding to one project in real images have essentially the same or similar brightness value, so if a distorted pixel can be picked out from the image. It can usually be replaced with the average value of neighbouring pixels.

Pre-processing is done to remove unwanted illumination effects such as non-uniform illumination, shadowing, aliasing and blurring, noise. It retains useful information like Facial features: eyes, nose, wrinkles, local shadowing, and shading. It involves the following steps,

Step 1: Gamma correction.

Step 2: Difference of Gaussian filtering.

Step 3: Contrast normalization.



Fig2: Stages of pre-processing chain

3.1.1.1 Gamma correction

It is a nonlinear gray-level transformation that replaces I gray-level with I^γ where γ is a user-defined parameter. This enhances the local dynamic range of the image in dark or shadowed regions while compressing it in bright regions and at highlights. A power law with exponent γ is used. Here we use $\gamma=0.2$.

3.1.1.2 Difference of Gaussian Filtering

Gamma correction does not eliminate the influence of overall intensity gradients such as shading effects. DoG filtering is a convenient way to achieve the resulting band pass behaviour. Fine details remain critically important for recognition so the inner (smaller) Gaussian is typically quite narrow ($\sigma_0 \leq 1$ pixel) while the outer one σ_1 might have of 2–4 pixels or more, depending on the spatial frequency at which low frequency information becomes deceptive rather than informative.

3.1.1.3 Contrast Equalization

The final stage of our preprocessing chain rescales the image intensities to standardize a robust measure of overall contrast or intensity variation. It is important to use a robust estimator because the signal typically comprises of extreme values produced by highlights, small dark regions such as nostrils, garbage at the image borders, etc. One could use (for example) the median of the absolute value of the signal for this, but here we have preferred a simple and rapid approximation based on a two stage process as follows:

$$I(x,y) \rightarrow \frac{I(x,y)}{(\text{mean}(|I(x',y')|^2))^{1/a}}$$

$$I(x,y) \rightarrow \frac{I(x,y)}{(\text{mean}(\min(r, |I(x',y')|^a))^{1/a}}$$

Here, a is a strongly compressive exponent which reduces the influence of large values, r is a threshold used to truncate large values after the first phase of normalization, and the mean is over the whole By default we use $a= 0.1$ and $r= 10$. Here we use the hyperbolic tangent $I(x, y) \rightarrow \tau \tanh(I(x, y)/\tau)$, limiting I to the range $(-\tau, \tau)$.

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3.2 Robust Feature Extraction

Robustness is the quality of being able to withstand stresses, pressures, variations or changes in procedure or circumstance [5]. A system, organism or design may be said to be "robust" if it is capable of coping well with variations (sometimes unpredictable variations) in its operating environment with minimal damage, alteration or loss of functionality. Our algorithm strives to achieve this robustness by incorporating some worthy feature extraction techniques. Current feature sets offer quite good performance under illumination variations but there is still room for improvement.

3.2.1 Local Binary Pattern (LBP)

Local Binary Pattern was introduced by Timoojala. The standard version of the LBP of a pixel is created by thresholding the 3x3 neighborhood of each pixel value with the center pixel's value. Let g_c be the center pixel gray level and g_i ($i=0,1,..7$) be the gray level of each surrounding pixel "Fig. 3a" illustrates the basic LBP operations. If g_i is smaller than g_c , the binary result of the pixel is set to 0 otherwise set to 1. All the results are combined to get 8 bit value. The decimal value of the binary is the LBP feature.

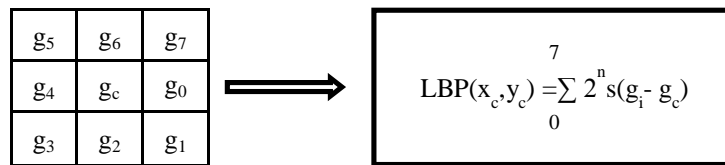


Fig3a: The LBP operator of pixel with 3x3 neighborhood

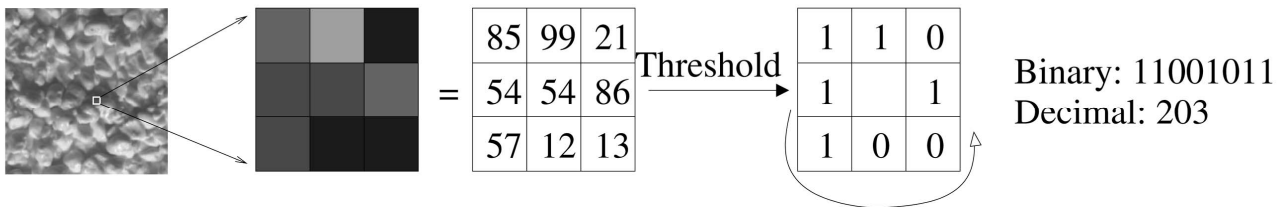


Fig3b: Illustration of basic LBP Operator

3.2.2 Local Ternary Pattern (LTP)

LTP consistently outperforms LBP by a small margin. LBPs have proven to be highly discriminative features for texture classification and they are illumination variation resistant in the sense that they are invariant to monotonic gray-level transformations. However because they threshold at exactly the value of the central pixel they tend to be sensitive to noise, particularly in near-uniform image regions, and to smooth weak illumination gradients. This technique extends LBP to 3-valued codes, in which gray-levels in a zone of width $\pm t$ around are quantized to zero, and the ones above this are quantized to +1 whereas the ones below it to -1. Coding way of LTP is shown in "Fig. 4".

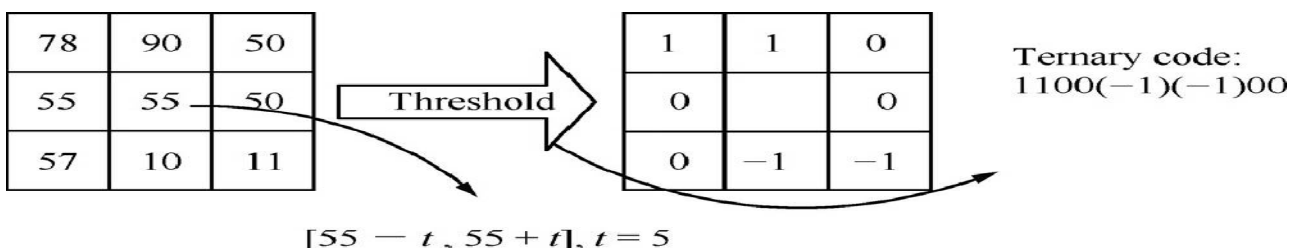


Fig4: Illustration of basic LTP Operator

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The calculation of the three values is shown as follows:

$$s'(u, i_c, t) = \begin{cases} 1, & u \geq i_c + t \\ 0, & |u - i_c| < t \\ -1, & u \leq i_c - t \end{cases}$$

Here t is a user-specified threshold, which is set as 5.

3.2.3 Gabor filter

Wavelet transform could extract both the time (spatial) and frequency information from a given signal, and the tunable kernel size allows it to perform multi-resolution analysis. Among various wavelet bases, Gabor functions provide the optimal resolution in both the time (spatial) and frequency domains, and the Gabor wavelet transform seems to be the optimal basis to extract local features for several reasons:

- Biological motivation: The simple cells of the visual cortex of mammalian brains are best modeled as a family of self-similar 2D Gabor wavelets.
- Mathematical and empirical motivation: Gabor wavelet transform has both the multi-resolution and multi-orientation properties and are optimal for measuring local spatial frequencies. Besides, it has been found to yield distortion tolerance space for pattern recognition tasks.

3.2.4 Elongated Multi Block Local Ternary Pattern (EMB-LTP)

It is a novel variant of LBP accommodating the main merits of three variants of LBP namely Elongated LBP, Multi Block LBP and LTP. In this model, the computation is done based on average values of block subregions rather than individual pixels. In this method, a multi scale block pattern is obtained considering the elliptic neighbourhood definition then a ternary pattern is obtained from the multi-scale block pattern.

We split each ternary into its positive and negative parts to reduce the running time. Then we obtain the positive and the negative EMB-LTP codes. In the positive parts, the negative values were set to zero, while the others remained. In the negative parts, both the positive values of LTP and zero ones were set to zero, and the negative values were set to one. In this way, the MBLTP code is divided into two LBP codes.

The EMB-LTP operator encodes not only the microstructure but also the anisotropic structural information and macrostructure of face image patterns, and hence provides a more complete image representation.

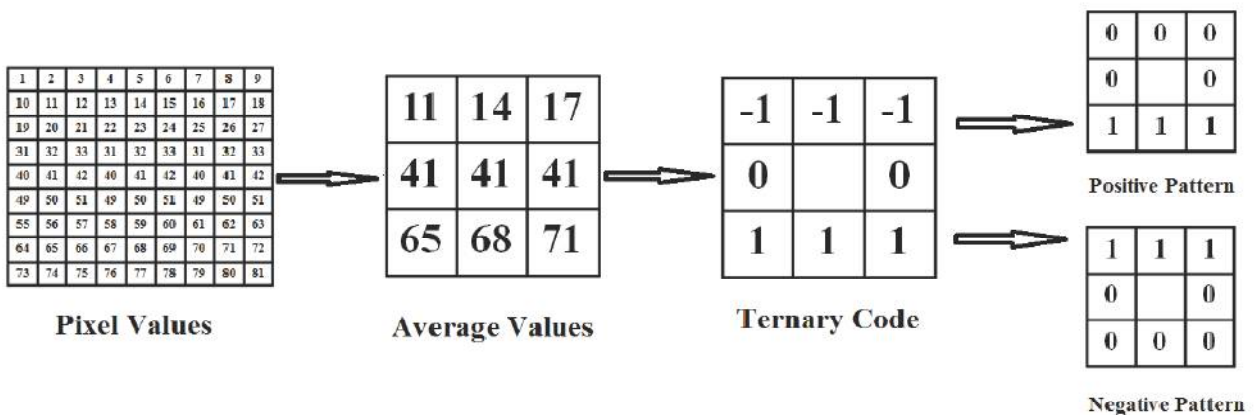


Fig5: Illustration of EMB LTP Operator

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3.3 Principal Component Analysis (PCA)

PCA is a statistical approach used for reducing the number of variables in face recognition. In PCA, every image in the training set is represented as a linear combination of weighted eigenvectors called Eigenfaces. These eigenvectors are obtained from covariance matrix of a training image set. The weights are calculated out after selecting a set of most relevant Eigenfaces. Recognition is performed by projecting a test image onto the subspace spanned by the Eigenfaces to find the set of weights that describe the contribution of each vector. Next classification is done by measuring minimum Euclidean distance. The advantage of this approach over other face recognition systems is in its simplicity, speed and insensitivity to small or gradual changes on the face.

IV. RESULTS AND DISCUSSION

In this section we present Matlab simulation results of our method. The database used to test our proposed method is extended Yale B database that has images of human faces with different subjects under varying illumination conditions and facial expressions. Table 1 gives the default values of the parameters used in the algorithm.

Table 1: Default Parameter Values

Procedure	Parameter	Value
Gamma Correction	Γ	0.2
DoG Filtering	σ_0	1
	σ_1	2
Contrast Equalisation	α	0.1
	r	10
LTP	t	5

Fig.6 gives the output of the face after being subjected to different stages of our method i.e. after preprocessing, feature extraction using operators such as LBP, LTP, Gabor filter, EMB-LTP. Finally the test image is checked with the training set to identify a match using PCA.

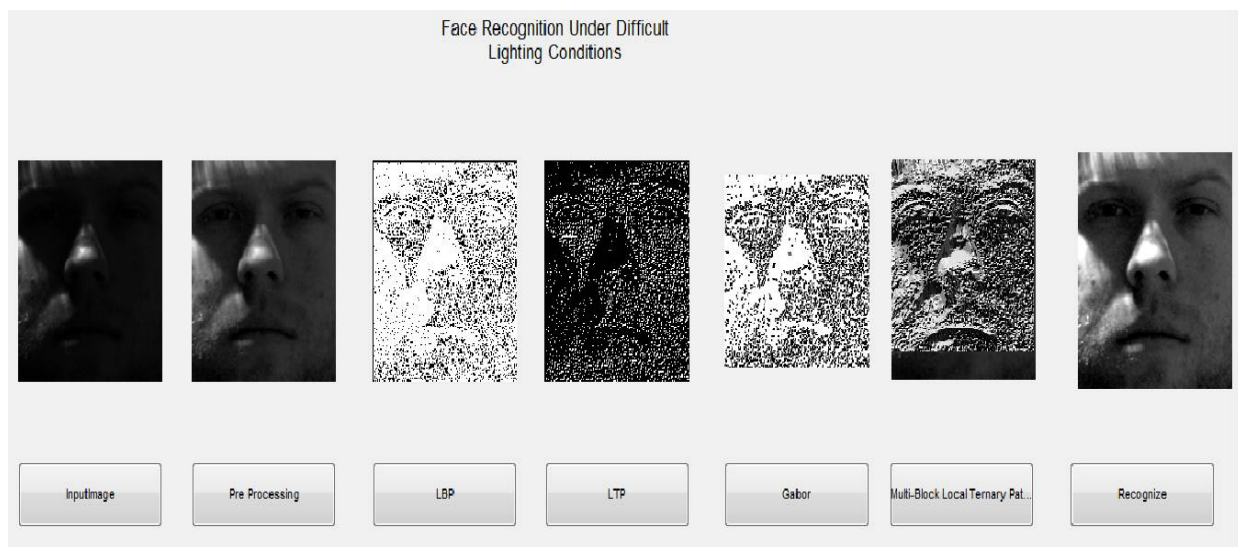


Fig6: Output of Test Image after each stage



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The metrics used to determine the performance of a face recognition algorithm are face verification rate and True rejection rate. They utilize terms such as true positive, true negative, false positive and false negative. The accuracy or the face verification rate is the measure of the test images being correctly identified if present in the training database. The rejection rate gives the information of the actual images of the database that are falsely rejected by the face recognition algorithm. The experiment is conducted over a number of subjects and Face Verification Rate and True Rejection Rate are calculated yielding a 90% Face Verification Rate and 4.4% True Rejection Rate.

V. CONCLUSION

This paper presents a novel face recognition algorithm that can surpass the challenges encountered due to varying illumination conditions because of its robust pre-processing techniques associated with a holistic measure of facial feature extracted by fusing multiple feature extraction techniques and thus compensating the shortcomings of one extraction technique with the advantage of another.

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