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RECOGNIZING PRE AND POST SURGERY FACES USING MULTIMODAL BIOMETRIC APPROACH

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Abstract: Variation in pose, expression, illumination, occlusion and aging are the major problem in face recognition and various algorithms have been proposed to handle these challenges. Except this new problem in face recognition is plastic surgery which changes facial features to large extent. Increasing popularity of plastic surgery and its effect on automatic face recognition has attracted attention from the research community. However, the variations introduced by plastic surgery remain difficult to be modeled by existing face recognition systems and degrade the performances of face recognition algorithm. This paper proposes a new face recognition system using multimodal biometric approach which extracts shape as well as texture feature from the facial image. Further to improve the recognition rate, cascade a periocular biometric.

Keywords: Face Recognition, Multimodal Biometric, Periocular Biometric, Plastic Surgery.

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INTRODUCTION

Recently, technology became available to allow verification of "true" individual identity. This technology is based in a field called "biometrics". Biometric access control are automated methods of verifying or recognizing the identity of a living person on the basis of some physiological characteristics, such as fingerprints or facial features, or some aspects of the person's behavior, like his/her handwriting style or keystroke patterns. Since biometric systems identify a person by biological characteristics, they are difficult to forge. Among the various biometric ID methods, the physiological methods (fingerprint, face, DNA) are more stable than methods in behavioral category (keystroke, voice print). The reason is that physiological features are often non-alterable except by severe injury. The behavioral patterns, on the other hand, may fluctuate due to stress, fatigue, or illness. However, behavioral IDs have the advantage of being no intrusiveness. People are more comfortable signing their names or speaking to a microphone than placing their eyes before a scanner or giving a drop of blood for DNA sequencing. Face recognition is one of the few biometric methods that possess the merits of both high accuracy and low intrusiveness. Also, it provides information about Age, gender, personal identity (physical structure), Mood and emotional state (facial expression) and Interest / attentional focus (direction of gaze). However, even after decades of research, face is still an active topic because of the variability observed in face due to illumination[11], pose, expression and occlusion[12]. A new challenge to face recognition is facial plastic surgery[8]. These surgeries alters the facial features to such an extent that human being often struggle to identify a person face after surgery. The Fig. 1 shows an example of the effect of plastic surgery on facial appearances.



Fig. 1 The effect of plastic surgery on facial appearances.

Due to advances in technology, affordability, and the speed with which these procedures can be performed, several people undergo plastic surgery for medical reasons and some choose cosmetic surgery to look younger or for better appearance. The procedures can significantly change the facial regions both locally and globally, altering the appearance, facial features and texture. Again, due to privacy issues, the surgical details of a particular individual are not

available and plastic surgery face database contains one pre-surgery image for training and a post-surgery image for testing. This further complicates feature extraction task in face recognition methods.

Also, each facial plastic surgery changes shape or texture of a particular face region[8]. It is very difficult to predict which features are invariant (a region without surgery effects) with unavailable surgery information. The difficulty is further supplemented, when an individual undergoes more than a surgery. The existing face recognition algorithm are good in extracting one of feature from an image i.e. either shape or texture.

The plastic surgery can also be misused by individuals who are trying to conceal their identity with the intent to commit fraud or evade law enforcement. Also this surgery allows the theft or terrorist to freely move around without any fear of being identified by any face recognition system. Again it might lead to rejection of genuine users.

Therefore the proposed method is used to recognition of facial images that have previously undergone some feature modifications through plastic surgery. Under this creating the plastic surgery face database is an important task which contains one pre-surgery image for training and a post-surgery image for testing. In the real world, it is difficult to isolate individuals who have undergone plastic surgery and use special mechanism to recognize them. Therefore, face recognition algorithms should be robust to variations introduced by plastic surgery even in general operating environments. Considering such generality of face recognition, the second non- surgery face database is prepared by appending the plastic surgery face database. Images in the plastic surgery face database are collected from different sources on internet and have noise and irregularities. So some preprocessing done on those images. Then the shape and texture features are extracted from the images and on the basis of that features the classification is done.

Face recognition across plastic surgery is further become difficult, when a person undergoes more than a surgery. Cascading another biometrics [5] information will help in reducing false positives and false negatives. Obtaining other biometrics information for available plastic surgery face database is again an exigent task. Part of face image which can serve as another source of biometrics is more attractive. Recent work proves that the periocular regions are even invariant to age. These regions are influenced by spectacles, head angle, hair and expression. Thus, multi-modal biometrics can surpass the limitations encountered by uni-modal biometric system [5]. Fusing periocular region features with facial features can effectively overcome plastic surgery hurdle in face recognition. Again it have the advantage that the user

doesn't have to provide two biometric, since the periocular region is obtain from face image as well as it is not required to processed all the biometric every times i.e. only when face recognition using proposed method fails to match then periocular biometric is used.

The rest of the paper is organized as follows. Section II briefly discusses a few related works from the literature. A brief discussion of proposed approach for face matching with preprocessing, feature extraction, classification and periocular biometrics is provided in Section III. Finally, this paper is concluded with section IV.

I. EXISTING APPROACH

Gaurav Aggarwal, Soma Biswas, Patrick J. Flynn and Kevin W. Bowyer[3], proposed a novel approach to address the challenges involved in automatic matching of faces across plastic surgery variations. In the proposed formulation, they proposed a part-wise sparse representation Approach combined with the popular sparse representation to address the challenge of plastic surgery variations and utilizes images from sequestered non-gallery subjects with similar local facial characteristics to fulfill this requirement. They stated that this sparse representation approach also used for several other biometrics and computer vision problems. One limitation of sparsity-based biometric recognition is, it requires several images per subject in the gallery.

K. R. Singh, Roshni S Khedgaonkar, Swati P Gawande [4], proposed a new approach to find the nearness between the pre plastic surgical face to the post plastic surgical face. They develop a classifier for facial images that have previously undergone some feature modifications through plastic surgery based on near set theory. Their work concerned only geometrically obtained feature values and their approximation using near sets. Once the features will be extracted a feature database will be formed. Using this feature values near set theory provides a method to establish resemblance between objects contained in a disjoint set, that is it provides a formal basis for observational comparison and classification of the objects. One limitation to this approach is, it will recognize the face only after local plastic surgery, but not work in the presence of global plastic surgery.

Himanshu S. Bhatt, Samarth Bharadwaj, Richa Singh, and Mayank Vatsa [1], proposed a multiobjective evolutionary granular algorithm to match face images before and after plastic surgery. The algorithm first generates non-disjoint face granules at multiple levels of granularity. The first level of granularity processes the image with Gaussian and Laplacian operators to assimilate information from multiresolution image pyramids. The second level of granularity tessellates the image into horizontal and vertical face granules of varying size and

information content. The third level of granularity extracts discriminating information from local facial regions. After feature is extracted from that face granules by SIFT and EUCLBP algorithm. Then Multiobjective Evolutionary Approach is use to optimization of weight. Decision is take place on the basis of weight.

II. PROPOSED APPROACH

This paper proposes a new multimodal biometric using face and periocular biometric for the recognition of face invariant to plastic surgery. This method makes the use of different features from face and periocular region to match face images before and after plastic surgery. The block diagram of propose method is shown in Fig. 2. Feature is extracted from both face and periocular region with the help of local binary pattern and then dimension reduction is done with the help of PCA. Then for classification Euclidian distance is used. If face is not match, then periocular biometric is performed for face recognition under plastic surgery. The flowchart contains the following steps: Data Collection, Preprocessing, Feature Extraction, Classification, Periocular biometric.

Data collection

The As the plastic surgey face database is not available, data required for face recognition across plastic surgery is collected from different sources on internet. This images have noise and irregularities. So some preprocessing like image contrast maximization, filtering is done. Using these images, the plastic surgery face database is created. This plastic surgery face database contains one pre- and post-surgery face image with frontal pose, proper illumination, and neutral expression. The database consists of different types of facial plastic surgery cases such as rhinoplasty (nose surgery), blepharoplasty (eyelid surgery), brow lift, skin peeling, and rhytidectomy (face lift). Face image from this database is used as a input to the proposed face recognition system.

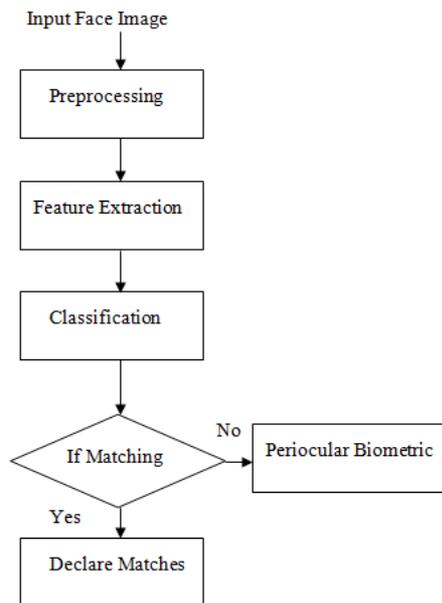


Fig. 2 The block diagram of proposed method

Preprocessing

The feature and information of face image should not be altered by local changes due to noise and illumination error. Hence to satisfy the environmental conditions, preprocessing of the raw data is highly important.

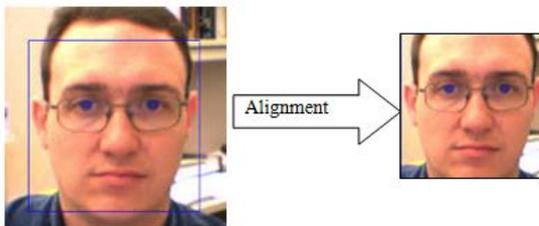


Fig. 3 Alignment process

Image capturing is a random process. The resolution of various image capturing devices may not be the same. This results in different resolution of the captured images. For accurate comparison of the features and to reduce the computational effort needed for processing, all the images should be scaled to a uniform size. So, first step of preprocessing is normalization in which face images are geometrically normalized and size of each image is

uniform. Further alignment is done in which we achieve more accurate localization of face. The result of alignment process is shown in Fig. 3.

1. Feature Extraction

Another phase in face recognition is feature extraction. This is phase where the system does the localizing of the characteristics of face component (i.e. eyes, mouth, nose etc) in an image. In other words, feature extraction is a step in face recognition where the system locates certain points on the face such as corner and centre of the eyes, tip of the nose, mouth, etc. it analyze spatial geometry of differentiate feature of a face. The result of this analyzing is a set of template generated for each face. The template consists of reduced set of data that represent the uniqueness of the face image. This proposed method uses the LBP for extraction of feature from face region.

In this, first extraction of shape free patch is done and then the features are extracted from the shape free patch with the help of local binary pattern operators. The sixty eight shape landmark points for the plastic surgery face database is manually selected for training purpose.

Local binary patterns[5] were introduced by Ojala as a fine scale texture descriptor. In its simplest form, an LBP description of a pixel is created by thresholding the values of the 3 * 3 neighborhood of the pixel against the central pixel and interpreting the result as a binary number. The process is illustrated in Fig. 4.

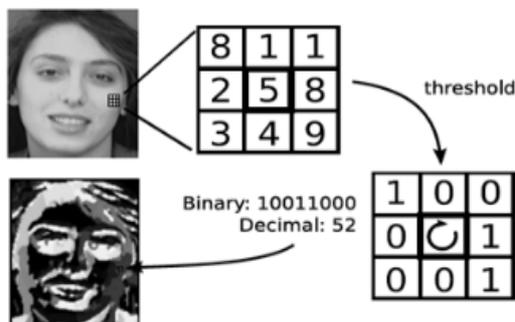


Fig. 4 The LBP operator thresholds each pixel against its neighboring pixels and interprets the result as a binary number. In the bottom image each gray-level value corresponds to a different local binary pattern.

Classification

Classification will be executed on the base of defined features i.e. it required some features such as density, texture or shape feature for classification of object. There are various classification techniques available, but I use Euclidean distance [5]. It is used as the classifier to identify which training set image belongs to the given test image. Classification is performed by comparing C from each training set image with the test image C_{test} using Euclidean distance, ϵ_i

$$\epsilon_i^2 = (|| C_{test} - C_i ||)^2$$

Where, C_i is a shape texture parameter of the i th face image in training set. Test image is classified as belonging to image i when minimum of ϵ_i is below some chosen threshold value θ . Threshold value, $\theta = 1/2 \max(|| C_j - C_i ||)$ where i and j are images from same class.

Periocular Biometrics

Eye lids, eye brow and eye surrounding area is called as periocular region[2] which is considered to be more discriminative in nature There is no database available with periocular region images. Only way to fetch this is using available face image[7]. Periocular biometric is a process in which the periocular region feature can be used for the classification. Periocular biometrics [7] is performed in three different ways such as overlapping, Non-overlapping and Strip [5]. All this three different types of periocular regions are obtained using four significant points in the eye region and lips are shown in Fig. 5. LBP is used for feature extraction and PCA is used for dimension reduction.



Fig. 5 Different types of periocular regions.

III. CONCLUSION

This paper presents an approach for the recognition of surgically alter human face. This paper proposes a multimodal biometric system which extracts features from face and periocular area

using local binary pattern operator. This clearly extract the shape and texture features which represents a face image in more meaningful way than any other feature extractor. The periocular biometric improving the overall system performance.

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