

Multi Instance Finger Knuckle Print for Secure Authentication

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Abstract:

Biometrics authentication is an effective method for automatically recognizing a person's identity. Most of the conventional biometrics modalities are used for security. These biometrics technologies provide proof of the physical and behavioral presence of a person for identification. Privacy and security are one of the most important challenges in biometrics traits. This paper proposed the secure user authentication using multi-instance as a means to improve the performance of finger knuckle print verification. A log-Gabor filter has been used to extract the image local orientation information, it represent the finger knuckle print features. Experiment is performed using the PolyU FKP database. Biometric performance using score level with simple sum rule outperform higher performance than using any single instance.

Keywords: FKP identification, Multi-instances; Log-Gabor, Score Level Fusion.

I. Introduction

Personal authentication is a very common issue to both academic research and industries. Because it has numerous applications about security for application of networking and mobility etc.; these applications require reliable authentication schemes to confirm the identity of an individual requesting their service for security. Usual personal authentication scheme using passwords and ID cards are commonly used to restrict access to a variety of systems and services. But these systems are helpless to attack and security can be easily breached. The need for reliable computerized user authentication techniques has been important for this reason [1]. Many researchers and academicians have presented the different biometrics traits like fingerprint, face, ear, iris, palm print, hand geometry, voice, and gait, signature etc. [2]. Among all the traits the hand based modalities like palmprint [4], hand geometry [5], hand vein [6], fingerprint [3], finger knuckle [10] and finger vein [13] create attentions for further enhancement. All these traits are highly accepted and user friendly for various security purposes. From the survey of different traits

it has been seen that researchers have less concentration on the finger knuckle print which actually provides high level security to identify for personal authentication. Finger knuckle print provides the image pattern of skin present on the back surface of finger. The outer surfaces of finger joints have more understandable line features and smaller area than other biometric traits. Due to this we have motivated for research and improve the performance on this biometric technique finger knuckle print (FKP). It refers to the image of the outer surface of the finger phalangeal joint [14]. Finger knuckle is contactless, user centric and open access control. As it is contactless hence no chance of proof of physical presence i.e. anti spoofing. So there are some advantages in security applications: (a) it is hard to be abraded since people hold stuffs with inner side of their hands. (b) Unlike fingerprint, there is no disgrace of criminal investigation associated with finger knuckle print. So FKP has a high user acceptance rate [9]. (c) People rarely leave FKP on stuff surface, creation the loss of private data less possible. Hence finger knuckle print is considered to be one of the most secure biometric techniques for personal authentication in future.

Multi-instance biometrics means the use of the same type of raw biometric sample and processing on multiple instances of similar body parts, (such as two fingers, or two irises) also been referred to as multi-unit systems in the literature [9]. These systems can be cost-effective since a single sensor is used to acquire the multi-unit data in a sequential fashion, and these systems generally do not necessitate the introduction of new sensors nor do they entail the development of new feature extraction and matching algorithms. Multi-instance systems are especially beneficial to users whose biometric traits cannot be reliably captured due to inherent problems. Due to this reason we take FKP with multi instance by fusing the data at score level using simple sum rule to improve the performance of biometrics.

II. Related Work

Lin Zhang et al.[4] proposed an effective FKP recognition scheme by extracting and assembling local and global features of FKP images. Specifically, the orientation information extracted by the Gabor filters is coded as the local feature. The proposed scheme exploits both local and global information for the FKP verification. The authors experimental results conducted on FKP database indicate that the proposed scheme could achieve much better performance in terms of EER and the decidability index than the other state-of-the-art competitors.

T.C. Faltemier et al. [6] proposed a multi-instance enrollment for face recognition as a means to improve the performance of 3D face recognition. The authors show that using multiple images to enroll a person in a gallery can improve the overall performance of a biometric system. The authors demonstrated that when using multiple images to enroll a person, sampling from different expressions improves performance over sampling only the same expression.

M. Vatsa, et al. [7] proposed a generalized biometric match score fusion framework using belief function theory. Multi- instance iris verification and multi-unit iris verification are used as the two cases. Experimental results show that the proposed fusion framework with PCR rule can effectively fuse the match scores even when the individual biometric classifiers provide highly conflicting match scores

Tobias Scheidat, et al. [8] proposed a fusion of two instances of the same semantic, where semantics are alternative handwritten contents such as numbers or sentences, in addition to commonly used signature. The fusion is carried out by the combination of the matching scores of two instances of one handwritten semantic. The authors demonstrated that when using three semantics and fusion strategies, improvements can be observed in comparison to the best individual results.

2 Multibiometrics Fusion

Fusion Categories:

In multibiometrics there are five categories under fusion ways [11]

- Single trait and multiple sensors.
- Multi biometric trait.
- Single trait and multiple units.
- Single trait and multiple classifiers.
- Single trait and multiple instances or multi-instance.

Fusion can be done at any four possible levels namely sensor level, feature extraction level, matching score level, and decision levels. Each level of fusion has its own advantages and disadvantages. In feature level fusion the problem is the choice of best classifier for high dimensional joint feature vectors. In matching score level fusion the biometric score into similar

domain we make use of normalization techniques. Due to less complexity matching score level fusion is used widely by the researchers.

In experiment, the data have been fused at score using different simple sum rule for two instances combination out of four fingers.

III. Proposed Work

To reduce the above problem the propose authentication has a scope to use database with multi instances and improve the quality of images using pre processing techniques so that it affect on the performance parameter EER. The proposed system focuses on personal authentication using finger knuckle print , including the specifically designed FKP data acquisition device , FKP feature extraction and pattern matching algorithms. The experiments are developed for personal authentication using DZhang FKP data base. FKP images were collected from 165 volunteers, including 125 males and 40 females. The database contains FKPs from two different fingers, right index, right middle. The DZhang database is available at the website of Biometrics Research Centre, the Hong Kong Polytechnic University.

A. Finger Knuckle Print Identification

Each finger has three joints .There are three bones in each finger called the proximal phalanx, the middle and distal phalanx. The first joint is where the finger joins the hand called the proximal phalanx. The second joint is the proximal interphalangeal joint. The usage of Finger knuckle images for personal identification has shown hopeful results and generated a lot of interest in biometrics [15]. Finger knuckles of the human hand are characterized by the creases on them. These creases differ from person to person. In the FKP identification system, after collecting the FKP images then apply preprocessing techniques on all the training images then extract the feature from the finger images[16].Knuckle crease patterns and stray marks as a means of photographic identification. Such features are unique and can use for identification.

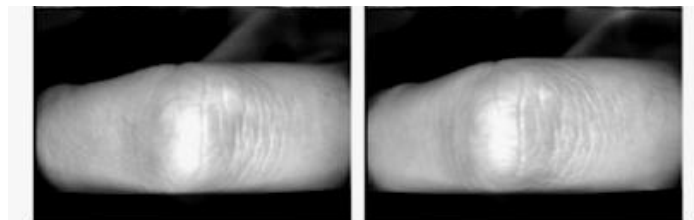


Fig. 1: Finger Knuckle Features

B. FKP Recognition System Design

The FKP recognition system is composed of an FKP image acquisition device and a data processing module. Below device in figure is composed of a finger bracket, a ring LED light source, a lens, a CCD camera and a frame grabber. The captured FKP image is inputted to the data processing module, which comprises three basic steps: ROI (region of interest) extraction, feature extraction and coding, and feature matching. Refer to Fig. 2; a basal block and a triangular block are used to fix the position of the finger joint. Fig. 3-a and 3-d show two sample images acquired by the developed device.

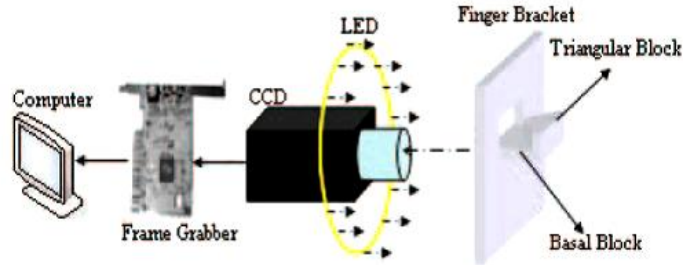


Fig. 2: FKP image acquisition device.

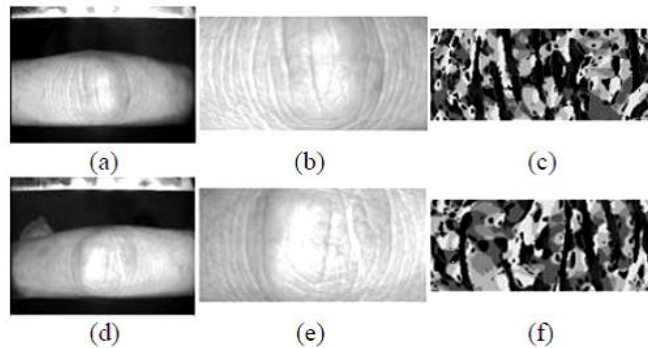


Fig. 3: (a) and (d) are two FKP images; (b) and (e) are the ROI images of (a) and (d); (c) and (f) are the competitive code maps generated

C. Region of Interest Extraction

It is necessary to construct a local coordinate system for each FKP image. With the help of this, region of interest can be cropped from the original captured image for reliable feature extraction and matching through different feature techniques. The complete steps for setting up such a coordinate system are as follows.

Step 1: determine the X-axis of the coordinate system. The bottom boundary of the finger can be easily extracted by a Sobel edge detector. This bottom boundary is nearly consistent to all FKP images because all the fingers are put perfectly on the basal block in data acquisition. By fitting this boundary as a straight line, the X-axis of the local coordinate system is determined.

Step 2: crop a sub-image SI. The left and right boundaries of SI are two fixed values evaluated empirically.

Step 3: Sobel edge detection. Apply a Sobel edge detection to SI to obtain the edge map EI.

Step 4: convex direction coding for EI. There will be ideal model for FKP curves. In this ideal model an FKP curve is either convex leftward or convex rightward. We code the pixels on convex leftward curves as 1 and pixels on convex rightward curves as -1 and the other pixels as 0.

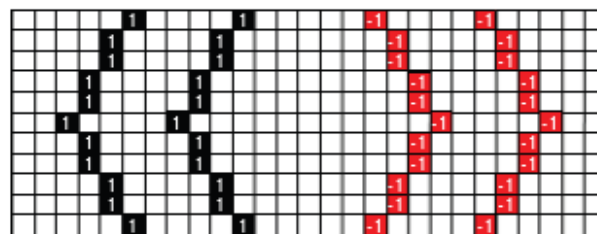


Fig. 4: Convex Direction Coding Scheme.

Step 5: now determine the Y-axis of the coordinate system. For an FKP image curves on the left part of phalangeal joint are mostly convex leftward and those on the right part are mostly convex rightward. In between curves in a small area around the phalangeal joint do not have obvious convex directions. Based on this observation, at a horizontal position x of an FKP image, we define the convexity magnitude.

Step 6: Crop the ROI image according to respective axis. As we have fixed the X-axis and Y-axis, the local coordinate system can then be determined and the ROI subimage ROI can be extracted with a fixed size. Fig. 3(b) and 3(e) presented two examples of the extracted ROI images.

D. FKP Feature Extraction

Log Gabor has been used as feature extraction algorithm; it eliminates the limitations in Gabor filters. Log-Gabor functions, by definition, always have no DC component, and, the transfer function of the log Gabor function has an extended tail at the high frequency end. Log-Gabor function proposed by Field [17], Field suggests that natural images are better coded by filters that have Gaussian transfer functions when viewed on the logarithmic frequency scale.

IV. Result and Discussion

These phase represent consequences comparisons of combining multi instance of finger knuckle print at score level fusion to measure the performance of multi instance system. In the experiment, performance is measured in terms of False Acceptance Rate (FAR in %) and corresponding Genuine Acceptance Rate (GAR in %). First the performance of a single instance biometric system is measured and later the results for multi instance biometric system fusion with simple sum rule at score level are evaluated. The results obtained from single instance biometric system are tabulated in Table-II.

Table II:Single Instance performance

FAR (%)	GAR (%)	
	Right- Index	Right-Middle
0.009	70.00	73.00
0.099	76.00	80.00
0.990	84.00	87.33

Next experimental results of the fusion of two instances with simple sum rule fusion at score level are shown. Table-III shows the fusion result of two instances. From Table-III it can be observed that the fusion of two in-stances finger at score level; simple sum rule has a significant improve score over the single instance. Whereas at score level; simple sum rule does not have performance improvement over a single instance. For this reason the performance will be like the best matcher.

Table-III: The Fusion Result of Two Instances

FAR (%)	GAR (%) Score level Simple
	Sum Rule Fusion
	Right-Index+Right-Middle
0.009	74.00
0.099	84.00
0.990	91.00

5. Conclusion

This paper presented a new identifier for personal authentication using multi instance finger knuckle print, which has distinctive line features. A cost-effective FKP system, including a novel image acquisition device and the associated data processing algorithms, was developed. Extensive experiments demonstrated the efficiency and effectiveness of the proposed technique. Compared with other existing finger back surface based systems, the proposed multi instance FKP authentication has merits of high level security with less EER, high speed, small size and cost-effective.

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