

Grenze International Journal of Engineering and Technology, July 2017

Finger Vein Biometrics for High Security

Shruthi G¹ and Sneha N Desai² ¹ Assistant Professor, DBIT/ECE, Bangalore, India ² PG Scholar, DBIT/ECE, Bangalore, India Email:shruthi.g04@gmail.com Email: snehanarayandesai@gmail.com

Abstract—Approach in finger vein identification to improve the performance of the system. For the verification of a person, the finger vein patterns are used which is more secure than other biometrics. Gabor filter is used in the proposed method for extracting vein features. Internal vein features can be extracted using this method. Depending on the angle, the vein patterns are tilted the orientation estimation is done. By using multi SVM matching technique, each person finger vein pattern is identified and is matched with that particular person.

Index Terms— Biometric; finger vein recognition; Gabor filter; Multi SVM classification; person identification.

I. INTRODUCTION

Compared to the other biometrics, Finger vein biometrics is easy to use and more comfortable. For the safety reason, the opposite biometrics, fingerprint, face reputation, voice popularity, Iris reputation are used. Earlier, the authentication was based on keys, password and magnetic card which are not safe as they will be stolen effortlessly or easily forgotten. To provide better protection, biometrics technology is implemented in a huge variety of system. Finger vein era has the exclusive vein patterns for the different person so it is more difficult to fool unlike the complex iris systems, high decision faux facial styles or false recordings.

When a person inserts a finger into an attester terminal containing a close to-infrared LED (mild- emitting diode) mild and a monochrome CCD (price-coupled tool) camera. Infrared LED light is absorbed by the haemoglobin inside the blood, which makes the vein gadget appear as a darkish sample of strains. The photo is recorded by the camera then the uncooked facts is digitized, certified and sent to a database of registered pictures. Finger is scanned for authentication purpose and the information is sent to the database. Blood vessel patterns are specific to every person, so blood vessel styles are nearly impossible to counterfeit because they're positioned below the skin's floor. Finger vein approach may be very secure and provide excessive protection for the identification of the person.

II. PROPOSED METHODS

The proposed method block diagram is shown in Fig.1 and each block is explained below.

Grenze ID: 01.GIJET.3.3.13 © Grenze Scientific Society, 2017

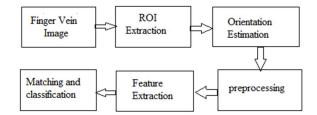


Fig.1 Block diagram of the proposed system

A. Finger vein image

Information acquisition involves the collection of pics to shape a database from the people. The infrared light is passed through the finger and finger vein images are taken and the haemoglobin present in the blood absorbs the radiation and givens the dark pattern due to absorption. To obtain finger vein pic, unique tool is used acquiring the pics of the finger vein without being stricken by ambient temperature.



Fig.2 Finger vein reader



Fig.3 captured image from the finger vein reader

B. ROI Extraction

In this technique, binarization and the brink mapping are used. In binarization, gray scale photo pixels are transformed in to either black or white pixels and threshold value is selected. Thresholding is the technique for image segmentation. Segmentation is the process of assigning each pixel in the supply image to two or greater lessons. If there are more than training then the same old result is several binary pix.

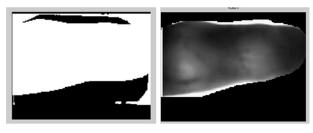
The operator use 3×3 kernels which is convolved with the unique image to calculate approximations of the derivatives - one for horizontal modifications, and one for vertical. A is defined as the supply photograph, and Gx and Gy are snap shots which comprise the horizontal and vertical by-product approximations respectively, the x-coordinate is described right here as increasing within the "right"-course, and the y-coordinate is increasing inside the "down"-course. The gradient magnitude is given by

$$\mathbf{G} = \sqrt{\mathbf{G}_x^2 + \mathbf{G}_y^2}$$

The direction of gradient can be calculated by

$$\Theta = \operatorname{atan2}(\mathbf{G}_y, \mathbf{G}_x)$$

where, Θ is 0 for a vertical edge which is lighter on the right side



2(a) Binarization image 2(b) edge mapping Fig.4 ROI extraction

C. Orientation Estimation

This is used for estimating the orientation of the photo. The fee of LRO at every pixel (i.e. the orientation image) is required as parametric enter to the filter. The approach is to decide LRO at a rectangular grid spaced (say) sixteen pixels apart, and achieve intermediate values through interpolation. The orientation field O is defined as a PxQ picture where in O (i,j) represents the local ridge orientation at pixel(i,j). There are some of strategies to calculate orientation fields; Local mean square estimation has been used. 1) The input image is divided into non-overlapping blocks

2) For each pixel, the x and y components of the gradient, G_x and G_y respectively, are calculated.

$$\begin{bmatrix} G_x = \delta p / \delta x \\ G_y = \delta p / \delta \\ G_x = \delta p / \delta \end{bmatrix}^T = \begin{bmatrix} \sum_w (G_x^2 - G_y^2) \\ W \end{bmatrix}^T 2G_x G_y]^T$$

The average gradient ϕ direction and dominant local orientation for the block are given by

$$\frac{1}{2} tan^{-1} \frac{\sum_{w} 2G_{x} G_{y}}{\sum_{w} (G_{x}^{2} - G_{y}^{2})}$$
$$(i, j) = \phi + \Pi/2$$

Additional low pass filtering is done in order to eliminate the wrongly estimated ridge.

D. Pre-Processing

This is the Image enhancement operation which improves the satisfactory of the picture and picture enhancement strategies can be grouped as either subjective enhancement or goal enhancement. Adaptive histogram equalization is a laptop picture processing method used to enhance comparison in pix. Evaluation constrained AHE (CLAHE) is different from ordinary adaptive histogram equalization in its contrast limiting. CLAHE was advanced to save the over amplification of noise that adaptive histogram equalization. Median filtering could be very extensively used in digital photograph processing because, below sure conditions, it preserves edges at the same time as getting rid of noise.



Fig.5 Enchanced image

E. Feature Extraction

A Gabor filter out is a linear clear out whose impulse response is described by harmonic feature increased by way of a Gaussian feature. The filters are convolved with the signal, resulting in a so-referred to as Gabor space. The Gabor space may be very beneficial in e.g., image processing packages inclusive of iris reputation and finger vein reputation.

A Gabor clear out may be considered as a sinusoidal plane of particular frequency and orientation, modulated by means of a Gaussian envelope.

 $\begin{aligned} \mathbf{G}(\mathbf{x},\mathbf{y}) &= s(\mathbf{x},\mathbf{y}) \ g(\mathbf{x},\mathbf{y}) \\ \text{where } s(\mathbf{x},\mathbf{y}) \ \text{is complex sinusoid and} \ g(\mathbf{x},\mathbf{y}) \ \text{is 2D gaussian envelope} \\ s(\mathbf{x},\mathbf{y}) &= \exp\left[-j2\pi \left(\mu_0 x + \upsilon_0 y\right)\right]. \\ g(\mathbf{x},\mathbf{y}) &= \frac{1}{\sqrt{2\pi\sigma_x\sigma_y}} \exp\left[-\frac{1}{2}\left(\frac{x^2}{\sigma_x^2} + \frac{y^2}{\sigma_y^2}\right)\right] \end{aligned}$

 σ_x and σ_y characterize the spatial extent and bandwidth of along the respective axes, u_o and v_o are the shifting frequency parameters in the frequency domain. Using G(x, y) as the mother wavelet, the Gabor wavelets, a class of self-similar functions can be obtained by appropriate dilations and rotations of G(x, y) through: $G_{m,n}(x, y) = a^{-m}G(x', y')$, where $x' = a^{-m}(xcos\theta + ysin\theta) = a^{-m}(xsin\theta + ycos\theta)$, $y' = a^{-m}(xsin\theta + ycos\theta)$, a>1, $\theta = \frac{n\pi}{0}$, $m = 1 \dots S$ $n = 1 \dots 0$. 0 indicates the number of orientations, S the number of scales in the multi resolution decomposition and a is the scaling factor between different scales. These parameters can be set according to reduce the redundant information (caused by the Non orthogonality of the Gabor wavelets) in the filtered images.

Or in polar coordinates,

Magnitude
$$\left(\widehat{g}(u,v)\right) = \frac{k}{ab} \exp\left(-\pi \left(\frac{(u-u_o)_r^2}{a^2} + \frac{(v-v_o)_r^2}{b^2}\right)\right)$$

Phase $\left(\widehat{g}(u,v)\right) = -2\pi \left(x_o(u-u_o) + y_o(v-v_o)\right) + P$



Fig.6 feature extraction using gabor filter

F. Matching and Classification

Vein matching, also called vascular generation, is a way of biometric identity through the evaluation of the patterns of blood vessels visible from the floor of the pores and skin. There are fundamental forms of finger vein matching techniques: graph based and trivialities primarily based. For contemporary embedded finger vein recognition structures, the trivia-based matching is popular This approach has excessive processing velocity and robustness to rotation and partial prints.

Alignment-primarily based matching algorithms take use of the shape of the ridge linked to trivia. This could improve the system accuracy. However, this approach effects in a larger template size because the related ridges for each minutia must be stored. Some different researches integrate the neighborhood and global structures.

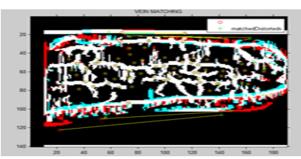


Fig.7 Matching and classification using multi-SVM

III. EXPERIMANTAL RESULTS

In this paper the 10 images are stored in the database. For each image the following step are carried out to find that the image belongs to which person. The features extracted from finger vein images are already stored in a database. The features of the input image are matched with all the extracted veins in the database to check whether the input image is matched with any one of the extracted veins. If the input image is matched with any one of the extracted veins. If the input image is matched with any one of the extracted veins, the message box will be opened and display "vein matched". If the input image is not matched with any one of the extracted veins, the message box will be opened and display "vein not matched". Then the vein matching score is calculated and it also determines to which person the vein patterns belongs which is stored in the database. The proposed method the matching percentage is 0.9534% and the vein belongs to the first person. If the matching score is more then 0.50% then the vein is matched or else the vein is not matched.

IV. CONCLUSIONS

Fully automated finger image matching framework is designed by simultaneously utilizing the finger surface and finger subsurface features. New algorithm is presented for the finger vein identification which is more reliable to extract the finger vein shape features and achieve much higher accuracy than the previous approaches.

Our finger-vein matching scheme works effectively in realistic scenarios and leads to a more accurate performance, as demonstrated from the experimental results. For the identification of low resolution finger surface texture images, an automated approach is examined. This investigation and the obtained results are significant as they point toward the utility of touch less images acquired from the webcam for personal identification and its extension for other utilities such as mobile phones, surveillance cameras, and laptops.

REFERENCES

- [1] D.Wu and S.-H. Yet, "Driver identification using finger-vein patterns with Radon transform and neural network," Expert Sys. And Appl., vol. 36, no. 3, pp. 5793–5799, Apr. 2009.
- [2] Z. Zhang, S. Ma, and X. Han, "Multiscale feature extraction of finger vein patterns based on curvelets and local interconnection structure neural network," in Proc. ICPR, Hong Kong, 2006, pp. 145–148.
- [3] N. Miura, A. Nagasaka, and T. Miyatake, "Feature extraction of finger vein patterns based on repeated line tracking and its application to personal identification, "Mach. Vis. Appl., vol. 15, no. , pp. 194–203, Oct. 2004.
- [4] N. Miura, A. Nagasaka, and T. Miyatake, "Extraction of finger-vein patterns using maximum curvature points in image profiles," in Proc. IAPR Conf. Mach. Vis. Appl., Tsukuba Science City, Japan, May 2005, pp. 347–350.
- [5] M. Kono, H. Ueki, and S. Umemura, "Near-infrared finger vein patterns for personal identification," Appl. Opt., vol. 41, no. 35, pp. 7429–7436, Dec. 2002.
- [6] J. Hashimoto, "Finger vein authentication technology and its future," in VLSI Symp. Tech. Dig., Honolulu, HI, 2006, pp. 5–8.
- [7] D. Mulyono and H. S. Jinn, "A study of finger vein biometric for personal identification," in Proc. ISBAST, Islamabad, Pakistan, 2008, pp. 1–8.