

Partial Fingerprint Recognition of Feature Extraction and Improving Accelerated KAZE Feature Matching Algorithm



P.Gayathiri, M. Punithavalli

Abstract: Nowadays for Personal Identification Numbers (PIN) and passwords an efficient and secure alternative has been used which is called e-security. Due to the various computer frauds like identity theft and computer hacking there is a rapid increase in the financial losses of payment. In present day, the banking sector is accepting the payments through credit cards for every on-line as well as offline transaction which actually encourages the cashless payments. It will be the foremost convenient because of do on-line looking out, bills payments etc. Therefore, there is also increase in the fraud risks that area associated with the misuse of the credit cards as a result of technology development. So we go for biometric security to resolve this problem. The aim of our study is to given security through biometric technology of fingerprint recognition. Minutiae extraction and minutiae matching technique of AKAZE to boost the accuracy and high speed computation.

Keywords : AKAZE, : Biometric, Crossing Number, Fingerprint Matching, Fingerprint Extraction, SIFT

I. INTRODUCTION

Fingerprints are gaining popularity due to the truth it's far outstanding for every body and one need no longer to remember them as the password. Fingerprints are commonly with a person and one want no longer to maintain like cards (credit/debit) [1]. Biometric recognition refers physiological (face, iris, palm print, fingerprint) as well as behavioral (signature, gait) features that are known as biometric identifiers. Due to increasing fraud the banking and also the business institutions have significantly increased fingerprints recognition systems in various applications like e-commerce, banking, ATM transactions, government concerns as well as forensic department Fingerprint recognition is one of the automated method that is used for confirming or identifying the individual's identity by comparing the two fingerprints. Also, among the various biometric methods one of the best method is fingerprint recognition and also for computerized system's authentication it is the mostly used biometric solution.

In comparison to the behavioral and physiological features like palm vein, hearing, gait, voice, iris and face features the

most leading biometric feature that is being used is the fingerprint. A fingerprint recognition system contains two stages that are verification and identification. In the first stage, an entry fingerprint is compared with the enrollment fingerprint of one to many matching verify whether the same person fingerprint or different fingerprint. In the identification stage an entry fingerprint is compared with the database fingerprint of many to many matching if the fingerprint is genuine person or fraud person.

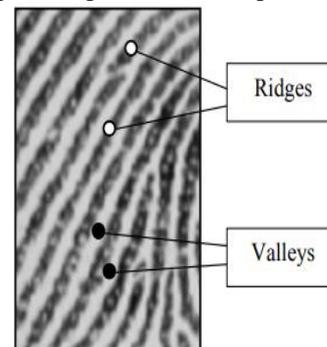


Fig.1 Fingerprint Ridges and Valleys

A fingerprint contains many dark patterns, valleys and ridges within the fingerprint is termed streak and therefore the light-weight region between the ranges is said because the furrow [2] in figure 1. There are some distinctive purposes of the fingerprint that are referred to as because the minutiae point.

Minutiae points of fingerprint patterns are bifurcation and termination, the bifurcation point contains the branches of ridges called (bifurcation) and termination point is a ridge end (termination) [3]. Though many sorts of minutiae will be typically classify the partial fingerprint problem and to improve with high accuracy. The most significant thing for fingerprint recognition are the smallest details within the partial fingerprint that contains minutiae check with specific points.

II. FINGER PRINT RECOGNITION

A. Matching Techniques

Minutiae-based total fingerprint matching stores an image in fixed points and a template is used that consist of the matching set of points and hence minutiae is entered.

- In a correlation-based totally overall matching of fingerprints, snapshots of fingerprints are overlaid and then correlation is calculated among equal pixels.

Manuscript published on 30 August 2019.

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Partial Fingerprint Recognition of Feature Extraction and Improving Accelerated KAZE Feature Matching Algorithm

- Ridge characteristic dependent fingerprint matching method is an associate degree revolutionary technique through which ridges are captured as a minutiae-based fingerprint capturing of an image of fingerprint in noise level is sufficient minutiae points.

B. Minutiae Features

Minutiae features as represented by figure 2, there are generally three levels of features that are extracted from a fingerprint image, that are:

- In level one option macro details such as singular points, pattern sort, and friction ridge flow are captured.
- In level two options ridge bifurcations and endings are referred.
- In level three options edge contour, pores, shape, width, ridge path deviation, and different details, together with inchoate scars, creases, and ridges are captured.

The level one option the fingerprint are categorized into key pattern sorts such as whorl, loop, or arch; whereas the level two as well as three options will be accustomed in establishing the singularity or individuality of fingerprint. If the resolution of fingerprint image is high, in that case it is easy to extract the high-level options.



Fig 2. Fingerprint Features

The partial fingerprint is the incomplete or the minor fingerprint part that contains only 25-30 minutiae points in fingerprint image. Generally the small scanners captures the partial fingerprint match among the images of two fingerprints. The small areas ranging from 1 "x1" to 0.42 "x0.42" can be detected by the miniaturized fingerprint sensors. However, only the partial fingerprints is captured by the fingerprint scanners that are having a detection area less than 0.5 "x 0.7" that is actually the fingerprint's average size.

The small number of minutiae in partial fingerprints remains a challenge for fingerprint matching.

C. Fingerprint Identification Applications:

- For detecting the criminals in criminality acts. It is among the main reasons that FBI in USA has developed this technology.
- It helps to improve the security for only authenticated person nor unauthenticated person.
- For automatically recognizing as well as billing from the debit card or credit card of registered user in the grocery store.

Current methods of partial fingerprint algorithms facing the following problems are:

- (i) Partial fingerprint trivia extraction algorithm necessary to decrease performance accuracy
- (ii) The correspondence of the minutiae deteriorates considerably with a number of minutiae also insufficient.
- (iii) Due to the inefficient and complex calculation time, the previous partial print work was unsuccessful. To

overcome these existing problem we have to improve the partial fingerprint through the feature extraction thinning method and Accelerated KAZE using hamming distance classifier is efficient way to improve the accuracy and the less computation time.

III. LITERATURE REVIEW

Ru Zhou [10] proposed Fingerprint identification the use of SIFT-based Minutiae Descriptors and the minutiae based fingerprint matching algorithms' overall performance degrades the accuracy compared with the existing algorithms Scale Invariant characteristic Transformation descriptors using SMD given high performance accuracy.

Helala AlShehri [7] proposes a singular fingerprint verification technique to resolve the trouble of move-matching. Fingerprints are captured from the equal finger using different sensors they range in scales and determination. To those troubles, the alternatives of a fingerprint are regionally extracted from it by using decomposing the fingerprint into overlapping home windows. Binary patterns and Scale Invariant Feature Transformation Descriptors using local ridge patterns. Descriptors can produce every rotation-invariant and scale-invariant LBP represent the vital systems of the fingerprints like edges SIFT has moreover been verified the descriptor due to the fact it is scale-invariant and rotation invariant.

Chikkerur [23] proposed novel graph based fingerprint matching algorithm. An invariant representation under rotation and translation called K-plet is used to define the local neighborhood for every minutiae. Every neighboring minutiae was represented using direct graph and every minutiae using a vertex which accumulated to encode a structurally local relationship for K-plet. A dynamic algorithm for programming was formed for local neighborhood matching. For consolidating every local match in fingerprints of two or more types, the algorithm used was "Coupled Breadth First Search Algorithm". A database of eight hundred images was used to evaluate the performance and analysis of matching algorithm.

Zhang Jie [12] presented pattern entropy based similarity measure for incomplete fingerprint recognition. The global orientation as well as local minutiae features are obtained and merged for constructing a unique rotation and scale invariant feature. The pattern entropy for similarity measure that eliminates most of the false matches and performs time complexity

Kai Cao [11] proposed a novel fingerprint classification algorithm orientation model is to improve the fingerprint extraction using Complex filter fingerprint ridge and singularities. A heuristic classifier that's powerful to intra-classification variability's and inter-category variability's to gain a higher accuracy.

Tsai-Yang [22] represented a new minutiae-based partial fingerprint matching that uses Support Vector Machine (SVM). They have designed an SVM to classify the partial fingerprint with the input (query) and (registered) fingerprints square measure at the beginning matched per their minutiae point's challenges of minutiae-based partial fingerprint matching square measure

- The problem is poor quality fingerprints.
- They need for additional features (than minutiae) to improve the accuracy.
- Global minutiae matching square measure neither robust nor efficient to handling distortion.

Wonjune [3] developed the Partial Fingerprint Matching process from ridge shape points (RSFs) in small ridge segments. In the matching degree of minutiae the appropriate minutia pair was decided by evaluating the nearest RSF and side by side with each minutia. During the next stage of matching the ridge operation, RSF increases matching accuracy.

Paul W.H [8] Proposed form context descriptors for dot pattern matching. The modified form matching context using certain contextual information can provide a boost for fingerprint matching when compared to the original image. To minimize calculations at the pre-processing step, what is called elliptical region filtering is turning off fake little things before matching.

Milstein [25] proposed an algorithm of fingerprint attention for partial and overall fingerprints. The Space Frequency Transformation Algorithm and Line Scanning Algorithm were developed to check partial fingerprints and reduce the time needed to assess full fingerprints. Basically, most of the advantages of the core factor of the algorithm are given excessive accuracy in partial fingerprints. At the moment the important weakness of the algorithm being developed is the lack of an initial form of small things that is validated.

IV. STATEMENT OF THE PROBLEM

Partial fingerprints for technology development of e-commerce transactions and banking sector have problems in identifying the real user. The efficient partial fingerprints use makes it all possible for using the technology like USB drives, Digital security credit cards and many others. On the other hand they should be used through a ways that can progressively authenticate them.

V. METHODOLOGY

A. Data collection

As there the partial fingerprint database is not available publically, therefore, proposed matching algorithm datasets performance of partial fingerprints is evaluated through the FVC2000 DB1 images simulation from optical sensor contains 800 images in 388 dpi.

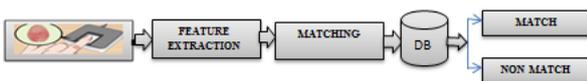


Fig. 3: Architecture diagram

B. Minutiae Extraction

In feature extraction minutia point can be of either type be a termination type or bifurcation type. For the representation of fingerprints, this technology is being used world widely and also it is having very distinctive configuration. And also this technique is having a small template size. Furthermore, if the minutiae points of two fingerprints have matched then only

system confirms that there is a match. Presently, among the various available products that uses fingerprint recognition the minutiae based fingerprint methods proved to be act as their backbone. The complex issues regarding the fingerprint recognition extraction and algorithm matching has been reduced through the minutiae feature fingerprint representation techniques [6]. The crossing number method of ridge pixel equation:

$$CN = 0.5 \sum_{i=1}^8 |P_i - P_{i+1}|$$

Pi represents pixel value in the neighborhood of P it encloses eight neighboring pixels in an anti-clockwise direction as in table 1.

Table 1 crossing number

P ₄	P ₃	P ₂
P ₅	P	P ₁
P ₆	P ₇	P ₈

Crossing Number Algorithm

Step 1: Find CN = 1 and CN = 3

Step 2: X and Y of minutia points

Step 3: direction θ of minutia

Step 4: detect the false minutia

IF feature = core point THEN

Set CN = 5

END IF

IF feature = delta point THEN

Set CN = 7

END IF

Step 5: Store information,

Minutia = [X, Y, CN, θ]

Crossing number computed with ridge pixel properties shown in table 2 specified Crossing number of 1 is ridge ending and crossing number of 3 specifies bifurcation. Parameter of x and y segmented the minutiae points in edge direction. The ridge detail will be categorized as a non-minutiae, bifurcation or ridge finishing. For instance a ridge detail along with crossing number of 1 corresponds to a ridge completing and 3 corresponds to a bifurcation [7] [8]. Their technique includes using a 3×3 window to test the nearby neighborhood of image's each ridge.

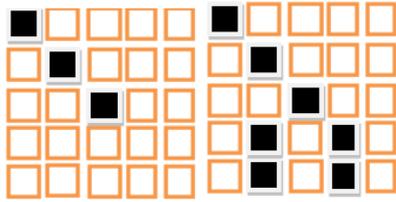
Table 2: Properties of the Crossing Number

CN	Property
4	Crossing Point
3	Bifurcation Point
2	Continuing Ridge Ending Point
1	Ridge Ending Point
0	Isolated Point

Furthermore, the ridge details are categorized as non-minutiae, bifurcation or ridge finishing. For instance, a ridge detail with a CN of one corresponds to a ridge completing and a CN of three corresponds to a bifurcation [7] [8].

Partial Fingerprint Recognition of Feature Extraction and Improving Accelerated KAZE Feature Matching Algorithm

Their technique includes using a 3x3 window to test the nearby neighborhood of every ridge in the image.



(a) CN=1 Ridge Pixel (b) CN=3 Bifurcation Pixel

1	0	0	0	0
p	1	0	0	0
0	0	1	0	0
0	0	0	0	0
0	0	0	0	0

1	0	0	0	0
0	1	0	0	0
0	0	1	0	0
0	1	0	1	0
0	1	0	1	0

(c) Ridge Pixel (d) Bifurcation Pixel

A detail is then categorized as a ridge finishing if it's far simply one near ridge element inside the window, and classified as a bifurcation if its far 3 near ridge pixels. Generally, a minutiae point which exists at the end of a ridge is known as termination minutia and the minutia points which exists at the cutting point or joining point of two ridges is known as bifurcation type of minutiae [5]. These minutiae points are the basis of the analysis in which are performed. In fig 3 terminations and bifurcations are shown with different colors. For termination type minutiae, red color is used while for bifurcation type minutiae point, pink color is used for indication and core point specified in green color. The issue regarding the point pattern matching to the complex issue of fingerprint identification is reduced by use of minutiae feature for fingerprint representation [4].



Fig3 (a) Original Image (b) Minutiae Extraction

C. POST-PROCESSING

A Fingerprint image is the place to begin of minutiae extraction (Lu et al 2002). Though its miles a miles described image, it's going to have deformations and cast minutiae that required to be filtered out. Since trivia are very not often adjacent, a set of rules might also abolish one in all adjoining minutiae. Scars, sweat or dirt may additionally motive abnormal minutiae that appear as false minutiae when obtaining the fingerprint image. Algorithms ought to hint any factors or styles that do not produce experience, which includes a spur on an island which maybe a false minutia. For numerous examples of fake minutia. In clockwise order,

interrupted ridges, forks, spurs, shape ladders, triangles and bridges are portrayed. Two very near strains with the same path create the interrupted ridges are removed.

D. AKAZE

An increased version of KAZE has also been created known as AKAZE which the creators claim has similar accuracy to KAZE while being extensively quicker. AKAZE function extractor and detector which gives a superb compromise among pace of calculation. A-KAZE became firstly proposed for object popularity as well as picture matching [12]. The key points of A-KAZE are extraordinarily one of a kind as well as are repeatable. For the descriptor generation and detection of A-KAZE key points following points are followed:

E. AKAZE Matching

To check whether the image match with the pair of enrollment image $Le [j]$ and template image in $Lt[i]$. If the matching between two pairs then points is calculated by hamming distance. This hamming calculates the distance threshold value in the following approaches.

Compare the template image function vector of enrollment image.

Calculate the hamming distances for each template function vector in equation

$$HD = \frac{1}{N} \sum_{j=1}^N X_j(XOR)Y_j$$

Let X and Y be the bit patterns for compare the Hamming Distance HD as in equation as defined as sum of exclusive OR between X and Y. N represents pattern's total bits number. Where X_i is the bit of $X_1, X_2, X_3, \dots, X_N$ and Y_i is the bit of $Y_1, Y_2, Y_3, \dots, Y_N$ of ones and zeroes. Find minimum distance of Le and Lt is calculated by means of Brute Force Method. if threshold greater than 0.5; contains matching otherwise nonmatching. Select the satisfactory pairs of matching key points arranged in ascending order in step with their convenience distance. Enroll distance HD template as the difference within the length of the rims with the aid of joining the image and the range of the enrollment image. The function is used for computing the location distance of provided key points.

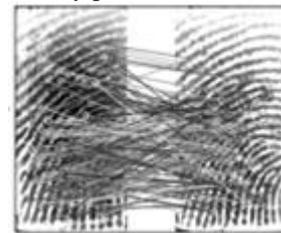


Fig 4 Matching point between two images

Algorithm 2 Matching key point pairs

- 1: Input: $LT [1,2,\dots,n]$, $LE[1,2,\dots,n]$
- 2: Output: Matched _Image [1, 2,...,n]
- 3: Conversion of Gray scale Image to Binary Image
4. Initialize
 $E[i] \leftarrow 0$; $i=1,\dots,n$
 $M[i] \leftarrow 1$; $i=1,\dots,n$
5. For $i=1$ to $n-1$ do
 For $j=i+1$ to m do

6. Hamming distance calculate the matching keypoints

$$HD = \frac{1}{N} \sum_{j=1}^N X_j(XOR)Y_j$$

```

if( Threshold<=0.5)
  Print("matching")
Else
  Print("non matching")
End
End
End
  
```

VI. EXPERIMENTAL RESULTS

A. A-KAZE using Hamming distance calculate the partial fingerprint match between the template and enrollment image pairs figure 5 shows the hamming distance performance accuracy is 98.5% of threshold value 0.5. Percentage of accreditation will be higher if considered as a complete data base. Experimental results shows in fingerprint recognition accuracy in FVC2002 dataset and for the separation point of 0.5 a false non-matching rate as well as a false matching rate of 0.095 and 0.075 respectively is obtained.

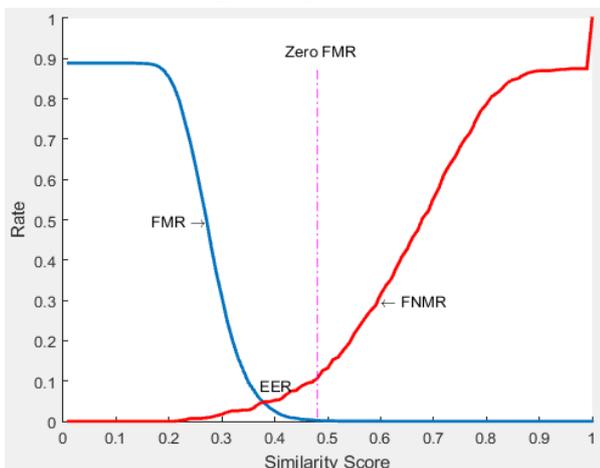


Figure 5 Matching Points between two images



Fig 6 Matching Time comparison of AKAZE and SIFT

Figure 6 represents the improvement of of proposed algorithm and existing algorithm Matching time and give the result interms of matching time accuracy.table shown matching time ,hamming distance between images and class labels of authorized person fingerprint should be matched or not.

Table 2 shows matcng tie and distance of akaze

Input Image	Matching Image	Matching Time	Hammin g Distance	Authenti cation
Img_101	1mg_109	0.00041	0.4327	Match

Img_105	Img_102	0.00032	0.7345	UnMatc h
Img_122	Img_143	0.00018	0.2643	Match
Img_110	Img_130	0.00072	0.1345	Match
Img_121	Img_113	0.00023	0.8344	UnMatc h
Img_151	Img_103	0.00042	0.3022	Match
Img_119	Img_111	0.00031	0.1222	Match
Img_112	Img_123	0.00023	0.8321	UnMatc h
Img_167	Img_134	0.00017	0.2322	Match
Img_156	Img_114	0.00025	0.1632	Match

VII. CONCLUSION

Overall the analysis of experiment partial fingerprint are degraded the performance accuracy in previous methods compared to existing method the proposed method feature extraction and minutiae matching AKAZE for faster and efficient partial fingerprint method.it has to perform higher accuracy using feature extraction thinning algorithm eliminate noise in fingerprint image and matching method of Accelerated AKAZE using hamming distance classifier improve the performance accuracy and less computation time

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Partial Fingerprint Recognition of Feature Extraction and Improving Accelerated KAZE Feature Matching Algorithm

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