Estimation of Modified RSA Cryptosystem with Hyper Image Encryption Algorithm

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Abstract

Generally, security, expediency and outlay are the three crucial factors influencing the espousal of biometrics. Seeing that, a newly scheme is proposed from the fused biometrics specifically fingerprint, finger vein and retina using Modified RSA (MDRSA), a public key cryptosystem used for both numeric and images. But the existing Hyper Image Encryption Algorithm (HIEA) is used only for images. Using MATLAB 2014, the performance is measured through GAR and FAR. Consequently, comparison of simulation results shows that MDRSA gives higher genuine rate of 95.3% and false rate of 0.01% which is reduced in case of HIEA, as it provides genuine rate of 92% and false rate of 0%.

Keywords: Biometric Authentication, Fused Biometrics, MDRSA, HIEA

1. Introduction

Currently, research works were carried in the field of single and multimodal biometric system. Multimodal biometric system has more advantages when compared to traditional single biometrics. Using various fusion algorithms and techniques, features are extracted and fused in case of multimodal biometric systems. Presently researchers proved that fused scores provides better discrimination than entity scores.

Author in[1] discussed fingerprint authentication technique based on minutiae feature extraction. To improve the matching performance, preprocessing and post processing stages were used in this system.

Author in[2] discussed about best accuracy from blood vessel extraction.

Author in[3] projected a multibiometric based on feature level fusion followed by fuzzy logic to improve security. The planned algorithm also gives the trade-off between accuracy and security.

Author in[4] developed a technique to extract retina blood vessels using curvelet transform.

To improve the recital of finger vein, a new approach is designed by Author in[5]. Preprocessing is performed for finger vein by image enhancement and feature extraction constructed by Gabor filter.

Author in[6] proposed to bind multiple biometrics with cryptography and fusion at the biometric level. The experiment was conducted on accuracy in terms of FAR and FRR at each model.

Author in[7] clearly explains authentication server attack with its issues and challenges.

Author in[8] anticipated a system to elevate the level of security.

Author in[9] presents an overview of various biometric template which overviews about security and privacy.

Author in[10] shows preventive issues in biometrics.

Author in[11] proves the limitations of feature level fusion in comparison with score level fusion for fingerprint and voice.

Author in[12] proposed a multimodal based fuzzy vault using iris, retina and finger vein. In this paper, discussion is carried about biometric security using feature level fusion.

2. Proposed Multimodal Biometric System

The courses involved in biometric system are shown in Figure 1.

Feature extraction is performed for the proposed

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biometrics namely fingerprint, retina and finger vein. These extracted features are encrypted using hyper image encryption algorithm and encrypted images are stored in a database for comparison and matching.

![Proposed multimodal biometric system.](image)

**2.1 Hyper Image Encryption Algorithm (HIEA)**

HIEA is a symmetric cryptographic algorithm which use block cipher instead of stream cipher. Three major steps involved in this algorithm are discussed below:

- Creation of transformation table.
- Encryption process.
- Decryption process.

**2.1.1 Transformation Table**

Creation of transformation table as in Figure 2.

![Clear representation of transformation table.](image)

**2.1.2 Encryption Algorithm**

Feistal structure of HIEA in encryption process is shown in Figure 3.

This is block based combination of permutation and followed by encryption algorithm and the encrypted data is stored in database. When user's comes under verification, key will generate. This key is used for both encryption and decryption process.

**2.2 Modified RSA (MDRSA)**

RSA is a public key cryptography where public key was used to encrypt and private key for decrypt the fused biometric images. In order to acquire fine eminence of the decrypted image, the amendment was made in decryption in RSA using symmetry properties of an algorithm.

MDRSA (Modified Rivest, Shamir and Adleman) was explained by the steps shown below:

**2.2.1 Key Making**

Figure 4 shows the key generation process in MDRSA.

![Key generation in MDRSA.](image)

**2.2.2 Encryption**

Given message ‘m’ is converted to ciphertext ‘C’ using the equation as shown in Figure 5.
2.2.3 Decryption

Ciphertext \( C \) is converted to plaintext \( m \) using public key ‘\( e \)’ and private key ‘\( d \)’. The steps involved in MDRSA decryption is shown in Figure 6.

\[
m = C^d \pmod n
\]

**2.3 Fused Biometric Techniques**

### 2.3.1 Finger Print Recognition

Most widely used biometric technology is fingerprint recognition technology as given in Figure 7.

**2.3.2 Retina Recognition**

Steps involved in retina recognition are explained as below:

- **Preprocessing** is the process which makes the input image suitable for further processing by image enhancement techniques.
- **Computation of Block Directions** determines the blood vessels area.
- Blood vessels in retina are extracted using **Segmentation**
- Extraction of blood vessels in finger vein is done using **Ridge Extraction**.
- **Minutia Extraction** determines the location and orientation of vessel bifurcation and vessel termination.
- **Post processing** is the process which eliminates extraneous minutia.

**2.3.3 Finger Vein Recognition**

Finger vein recognition in biometric recognition has excellent advantages like stability, uniqueness, immunity to counterfeiting, highly accurate etc. In proposed technique, pixels are identified using line-tracking scheme. The static position of the pixel is referred as current tracking point. The maximum curvature means locates the positions that acquire the maximum curvature from the image profile, and the profiles are acquired in different direction, while all points are extracted, they are connected and combined according to the rules. Hence vein pattern is obtained by connecting the positions with each other. The resulting binary mask is used to segment the ROI from the original finger vein image samples.

### 3. Simulation Results

The projected multimodal biometric scheme offers better performance in terms of GAR and FAR as discussed from the below results using MATLAB 2014. For instance, GAR and FAR for fused biometric gives better result when compared with single biometric. Designed fused biometric are trained with MDRSA and HIEA and similarly without MDRSA and HIEA.

Based on identical minutiae points in fingerprint, performance was deliberated for an identity using FAR of 10% and GAR of 72%.

For retina, 78% GAR and 6.74% FAR were measured.
The extracted finger vein gives genuine rate of 80% and false rate of 5.02% as in Figure 8.

The multimodal biometric was trained based on fused matrix values using correlation and it's GAR of 92% and FAR of 0%. The performance of finger print using HIEA has a GAR of 78% and FAR of 5.67%. Similarly the performance of Retina using HIEA has a GAR of 83% and FAR of 3.33%.

The performance of finger vein using HIEA has a GAR of 85% and FAR of 0.78% which was shown in Figure 9.

The fused matrix values of fingerprint using MDRSA gives 80% GAR and 3.25% FAR whereas for Retina, MDRSA gives GAR of 84.2% and FAR of 2.2%. The performance of finger vein using MDRSA has a GAR of 87.6% and FAR of 0.52%, as in Figure 10.

4. Conclusion

The primary benefit of feature level fusion is the detection of correlated feature values, generated by fused biometric which improves the recognition accuracy in terms of genuine rate and false rate. From the performance analysis of MDRSA in fused biometric, it is proved that GAR is increased to 95.3% and reduced FAR of 0.01%.
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