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Enhanced Contrast of Reconstructed Image for Image Secret Sharing Scheme Using Mathematical Morphology

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Abstract

Visual secret sharing (VSS) is one of the cryptographic techniques of Image secret sharing scheme (ISSS) that performs encoding of secret message image (text or picture) into noise like black and white images, which are called as shares. Shares are stacked together and secret message image is decoded using human visual system. One of the major drawbacks of this scheme is its poor contrast of the recovered image, which improves if computational device is available while decoding. In this paper, we propose to improve poor contrast of classical VSS schemes for text or alphanumeric secret messages and low entropy images. Initially, stacked image is binarized using dynamic threshold value. A mathematical morphological operation is applied on the stacked image to enhance contrast of the reconstructed image. Moreover, a method is proposed that allows the size of the structuring element to change according to the contrast and the size of a stacked image. We perform experiments for different types of VSS schemes, different share patterns, different share types (rectangle and circle), and low entropy images. Experimental results demonstrate the efficacy of the proposed scheme.

Keywords

Contrast, Mathematical Morphological Operator, Image Secret Sharing, Visual Secret Sharing

1. Introduction

An image secret sharing scheme (ISSS) is a cryptographic technique to hide secret message image into meaningless images, which reveal no information about the secret message; called as shares. Typically, the secret message image contains text or picture where text is short alphanumeric message or password. Super-

imposition of all shares reveals secret message. There are two major categories in ISSS: one is the visual secret sharing (VSS) scheme and the other is the polynomial-based ISSS (PISSS). VSS scheme is also known as visual cryptography scheme (VCS) [1]. Classical VSS reveals secret message without using any computational device and uses human visual system (HVS) only [1]-[8]. Typically, classical VSS schemes have recognizable reconstruction. Further, VSS schemes are also proposed for circular shares [6] [9] and cylinderical shares [13]. Contrarily, the PISSS can recover the secret image without any distortion, while it needs the computation [10].

Another category of VSS scheme proposed with two decoding options, where the secret image is revealed both by stacking the shares and by computation. This scheme is called as Two-in-one image secret sharing scheme (TiOISSS) [10] [11]. TiOISSS can decode secret images for preview by HVS when computational device is temporarily unavailable. When the computational device is available during the decoding scene, a high quality image is obtained for high-end applications. Though above-mentioned schemes give two decoding options, they suffer from pixel expansion and their lossless reconstruction is specific to the share generation schemes. Additionally, Wu *et al.* [12] showed decryptions using OR and XOR which is another example of two options decoding. Here, OR is used for HVS and XOR is applied for computational system. Furthermore, Viet *et al.* [16] introduced additional operation, called reversing, which can be used by participants in the reconstruction phase to enhance contrast. However, their scheme has many drawbacks including pixel expansion.

In this paper, we propose a mathematical morphology based scheme to achieve almost ideal contrast of classical VSS schemes for text message image. Here, image after stacking at receiver is considered as an input. Otsu's method is utilized to convert grayscale image into binary image. Further, image contrast enhances by mathematical morphological operation using appropriate size of square structuring element (SE). Notably, the proposed method computes size of SE from given input image itself.

Typically, for a given stacked text message image, impulse noise (pepper) is present. Briefly describing, few popular techniques to remove this impulse noise are isolated pixel removal technique, and adaptive median filter. Here, isolated pixel removal algorithm removes only isolated pixels, whereas probability of two or more pixels in a cluster is very high. Same way, adaptive median filter works well if the spatial density of the impulse noise is not very large. Hence, mathematical morphological opening operation is employed to remove impulse noise.

A possible application scenario of two options decoding as described in [12], is as follows. When computation resources are not available, stacking of shares decrypt the secret image though facing alignment issues. When computational device is available, contrast of the reconstructed image is enhanced up to almost ideal contrast by proposed method. Additionally, when computational resources are available during the decoding scene, a high quality image is obtained for high end applications [11]. It is noteworthy that the proposed scheme is independent of secret encoding and reconstruction, and works well for variety of image sharing schemes. The proposed scheme is like second step of TiOISSS [10]-[12] for VSS schemes for text secret message. Additionally, our scheme can also employ to low entropy secret images.

The rest of this paper is organized as follows: Section 2 describes the proposed scheme. Section 3 shows experimental results and further discussion of the proposed scheme, and in Section 4, we conclude this paper.

2. Proposed Scheme

Analogous to second step of TiOISSS, stacked image is considered as an input image. Stacked grayscale image has mainly foreground message and noisy background. Primarily, image is required to be converted into binary using optimum threshold value. Though histograms of the different input images represent bi-modal behavior, resultant threshold values vary in wide range. Hence, Otsu's method [17] evaluates optimum threshold value. Otsu's method is employed to perform automatically clustering-based thresholding. Subsequently, a morphological opening operation is applied. In mathematical morphology, opening is the dilation of the erosion of a set A by a structuring element B:

$$A \circ B = (A \ominus B) \oplus B \tag{1}$$

where $\ominus$ and $\oplus$ denote erosion and dilation, respectively.

Set erosion and dilation are formulated based on structuring elements (SEs). SE is a small set or subimage, used to probe an image under study for properties of interest. Here, we employ standard symmetric square binary SE of value 1. However, optimum SE width (W) is required to select for generalized solutions, as higher W of SE eliminates message content whereas lower W of SE keeps noise elements. Therefore, probability of

noise cluster is calculated for the optimum value of *W*. Here, input stacked result has typically white background (value 0) with pepper noise with black foreground message (value 1), whereas SE of width *W* eliminates $W \times W$ noise cluster. In stacked image, probability of black pixel is defined as,

$$P(B) = \text{Number of black pixels}/\text{Number of total pixels} \tag{2}$$

Whereas, probability of noise cluster of square *W* elements is,

$$P_n = P(B)^{W \times W} \tag{3}$$

While, possibility of occurrence of such noise cluster in $M \times N$ size image is,

$$Max_n = [M - W + 1] \times [N - W + 1] \tag{4}$$

The probability of noise cluster in computed image reduces exponentially with increasing dimension of SE. This is explained as follows. Here, as image size increases, Max_n increases which in turn, increases possibility of the noise cluster, whereas as *W* increases, P_n decreases which in turn decreases the probability of noise cluster. Hence, P_n and Max_n decide the W. Here, we consider high spatial resolution image, *i.e.* 1600×1200. For such high spatial resolution and typical lower value of *W*, Max_n approximates to $M \times N$ times noise region. As a result, Max_n is 1.92×10^{06}, which is defined as number of trials. Hence, P_n should be sufficiently small. Considering $W = 9$ and $P(B) = 0.75$ for VSS scheme, $P_n = 0.07585 \times 10^{-09}$. This means, probability of occurrence of one noise cluster requires 13.18×10^{09} trials ($1/P_n$), which is very high compared to the Max_n. As a result, there exists least probability that any noise cluster remains in an image after applying mathematical morphological opening operation. Lower value of *W* gives noisy reconstruction whereas higher value of *W* erodes secret message. To generalize the value of *W*,

$$\frac{1}{P_n} \gg Max_n$$

$$\frac{1}{P_n} = k \times Max_n$$

where *k* = number of attempts to generate noise cluster

Using Equation (3) and Equation (4),

$$\frac{1}{[P(B)]^{W \times W}} = k \times [M - W + 1] \times [N - W + 1] \tag{5}$$

If we solve Equation (5) for exact solution, we get e^{W^2} term. By expanding it as an exponential series, for a generalized solution, we can consider first three terms only as it results into a quartic function. Even though complex quartic function is considered, it gives an approximate solution only. Alternatively, *W* is neglected on the right hand side term as $M \gg W$ and $N \gg W$.

$$\frac{1}{[P(B)]^{W \times W}} = k \times [M + 1] \times [N + 1]$$

$$-(W^2) \ln P(B) = \ln \{ k \times [M + 1] \times [N + 1] \}$$

$$W = \left[\frac{\ln \{ k \times [M + 1] \times [N + 1] \}}{-\ln P(B)} \right]^{\frac{1}{2}} \tag{6}$$

Since *P(B)* is always < 1, denominator turns to positive value and as *W* is always positive integer, we do not consider the negative solution. Additionally, for float valued *W*, next odd value is selected. Here, small change in *P(B)* makes larger change in *W*, whereas numerator terms comparatively affects less. Logically, *P(B)*, probability of black pixels, is related to contrast, so as it changes, *W* changes rapidly. Whereas for same *P(B)* value, size of image is not much effected comparatively. Practically, for highly noisy images, *W* has large values. Very large value of *W* erodes secret message. Hence, cut-off value of *W* is set to 9. This cut-off can be increased

provided secret message width allows.

3. Experimental Results and Discussion

Experiments are conducted for different stacked messages [2]-[7] [12] [13] to demonstrate the feasibility of the proposed method. As a prerequisite step, we apply intensity thresholding using Otsu's algorithm followed by mathematical morphological opening operation. **Table 1** shows the threshold value obtained from Otsu algorithm, SE size calculated using proposed algorithm, and SE size considered for experiments for different type and different size of images. In case the calculated SE size results into a float number, our algorithm takes next odd natural number as SE size.

Here, **Figure 1(a)**, **Figure 1(b)**, and **Figure 1(d)** show secret image, result of stacking two shares, and result of stacking three shares of (2, 3) VSS scheme [2] whereas **Figure 1(c)** and **Figure 1(e)** represent the result of proposed method applied on (b) and (d) respectively. One secret image and stacked image of (2, 2) VSS scheme for sharing multiple secrets [3] and result of our scheme are presented in **Figure 2**. Likewise, secret image, stacked images, and resultant images of our algorithm are exhibited for two patterns of (2, 2) VSS [4] in **Figure 3**. Further, **Figure 4** demonstrates the results of (2, 2) and (3, 3) VSS using probabilistic method [5]. Additionally, outcome of (2, 2) VSS scheme using circular shares [6] for two secret images and VSS for cylindrical shares [13] are represented in **Figure 5** and **Figure 6** respectively.

Lastly, results showed for (2, 4) VSS using OR and XOR decryptions [12] where **Figures 7(a)-(d)** are secret image, result of stacking all four shares using OR, result of stacking all four share using XOR, and result of our method respectively. Experimental result for low entropy image [7] is presented in **Figure 8**, where (a), (b), and (c) show secret image, the stacked image, and the result for (b) using proposed scheme.

Results show that the proposed scheme works well for different types of VSS schemes, different share patterns, and different share types. Where **Figure 1** highlights (2, 3), (3, 3) VSS schemes, **Figure 4** shows effectiveness of the scheme for (2, 2) and (3, 3) VSS schemes. In [3], due to sharing multiple secrets, contrast of stacked image is very poor (1/15) which is improved to almost ideal as shown in **Figure 2**. Similarly, **Figure 3** represent efficacy of the method for different patterns. Besides rectangle shares, proposed scheme provides almost ideal reconstruction for circular shares as shown in **Figure 5**.

Table 1. Objective numerical parameters of **Figure 1** to 8 for k = 1000.

|	Stacked Image Size ($M \times N$)	Threshold Value	Structuring Element Size (Calculated)	Structuring Element Size (Experimental)	Probability of Black Pixels $P(B)$
Figure 1(b)	228×228	112	5.57	7	0.56
Figure 1(d)	228×228	112	5.55	7	0.56
Figure 2(b)	392×419	64	13.6	9	0.9
Figure 3(b)	272×272	107	5.67	7	0.57
Figure 3(d)	272×272	107	5.72	7	0.58
Figure 4(b)	240×120	112	6.11	7	0.63
Figure 4(d)	240×120	96	8.87	9	0.8
Figure 5(b)	270×270	64	5.42	7	0.54
Figure 5(e)	270×270	64	5.36	7	0.53
Figure 6(b)	360×255	80	4.98	5	0.48
Figure 7(b)	221×221	143	3.98	5	0.33
Figure 7(c)	221×221	46	4.28	5	0.38
Figure 8(b)	540×360	86	9.42	9	0.81
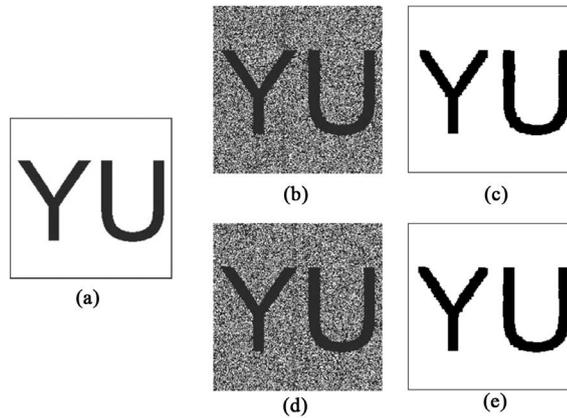
Figure 1. Using Chen [2] (2, 3) VSS (a) Secret image; (b) result of stacking two shares of Chen's (2, 3) VSS; (c) the result for (b) using our method; (d) result of stacking three shares of Chen's (2, 3) VSS; (e) the result for (d) using our method.
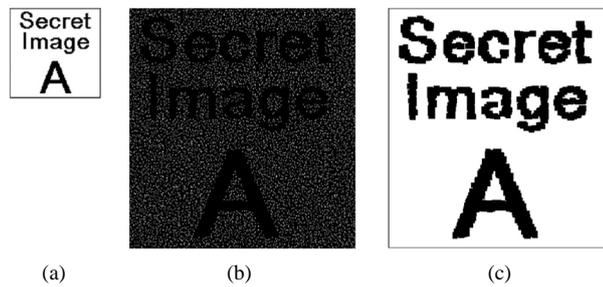

Figure 2. Using Feng [3] (2, 2) VSS (a) secret image; (b) the result of stacking for one secret message [3]; (c) the result for (b) using proposed method.
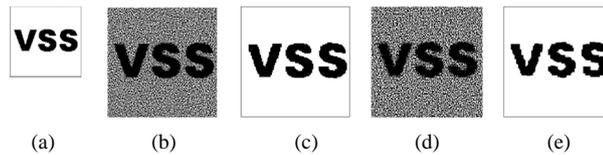

Figure 3. Using Yang [4] (2, 2) VSS (a) secret image; (b) the result of stacking for one pattern [4]; (c) the result for (b) using our method; (d) the result of stacking for another pattern [4]; (e) the result for (d) using our method.

Figure 4. Using Yang [5] ProbVSS scheme (a) secret image; (b) the result of stacking for Yang's (2, 2) ProbVSS; (c) the result for (b) using proposed method; (d) the result of stacking for Yang's (3, 3) ProbVSS; (e) proposed method result for (d).
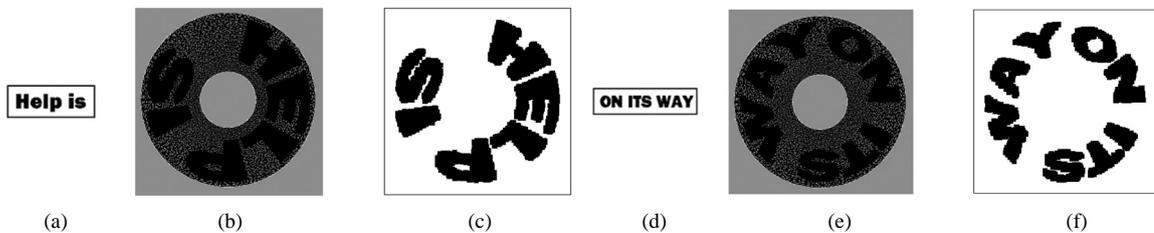

Figure 5. Using Shyu [6] multiple secret sharing in visual cryptography (a) secret image; (b) the result of stacking circle shares- $A \otimes B$ [6], (c) the result for (b) using our method; (d) secret image; (e) the result of stacking rotated circle share- $A^{120°} \otimes B$ [6]; (f) the result for (e) using our method.
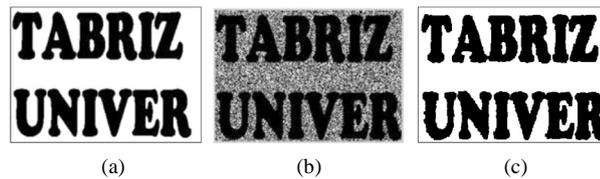
Figure 6. Using Salehi [13] VSS by cylindrical random grid (a) original secret image 1; (b) the result of stacking shares; (c) the result for (b) using proposed scheme.
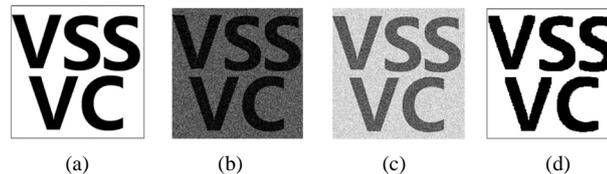

Figure 7. Using Wu [12] VSS with OR and XOR decryptions (a) secret image; (b) the result of stacking using Wu's OR decryption; (c) the result using Wu's XOR decryption; (d) the result for (b) using proposed scheme.
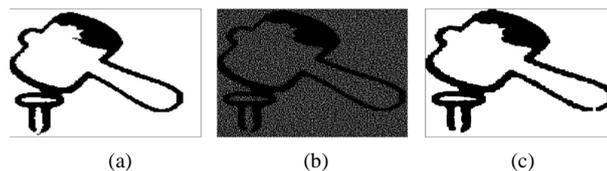

Figure 8. Using Lin [7] flip visual cryptography (a) secret image; (b) the result of stacking using Lins block-based expanded version of scheme 1; (c) the result for (b) using proposed scheme.

Additionally, we compare proposed method with the Wu *et al.* [12] method when all four shares are stacked. Here, our method gives improved performance compared with Wu *et al.* [12] when computational device is employed. Finally, experimental results prove that our method works well for low entropy images also. However, mathematical morphological operation smooths the stacked image. Though our scheme assures improved contrast for classical VSS, size of SE plays a significant role. For a stacked image, secret message width should be higher than experimental SE size else secret message erodes in first step itself. Moreover, we have also verified the proposed scheme for results of other schemes [8] [14] [15].

4. Conclusion

In this paper, we propose a generalized solution using mathematical morphological opening operation to remove noise clusters from the stacked message image of VSS. Stacked grayscale image is converted to binary image with distinct threshold value using Otsu's algorithm. Mathematical morphological opening operation is applied on resultant image. Width of square SE is formulated from the stacked image with higher cut-off value of 9. Here, contrast of the stacked image plays significance role compared to size. Even small difference in contrast value makes large variation in SE size whereas for the similar contrast values, SE varies little. The removal of noise from the image improves the contrast and reconstructs received image from recognizable to almost ideal contrast. Here, secret message will be eroded if the width of the secret message is less than the experimental SE. The proposed method works well for the application scenario of two options decoding techniques. Additionally, proposed scheme can be implemented for handheld devices like mobile phone.

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