QR-like Secret Data Hiding in Integer Wavelet Transform Domain Images

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Abstract Steganography is the process of hiding secret information in other multimedia files, like image, audio and video, such that any unauthorized person does not have any clue of the hidden secret data. Due to the recent computational advancements, there has been an increased risk in secure communication and thus, steganography has become one of the most researched domains in information security. In this paper, a novel method of steganography is approached where the cover image is scrambled using graph theory; then the secret data is embedded in the LH, HL and HH planes using LSB embedding of the cover image after applying Haar forward transform to get the embedded image. The secret data hidden in the cover image is first converted to a QR-like (Quick Response) image and then to be embedded in the cover media. The stego image is obtained after reverse Haar transform and descrambling of the embedded image. The proposed method when applied to a set of grayscale images, not only gave us stego images with greater embedding capacity but also high Peak Signal to Noise Ratio (PSNR) and a low Mean Square Error (MSE) and very appreciable Normalized Cross-Correlation (NCC) along with a good Structural Similarity Index Matrix (SSIM) when compared to the cover images.

Keywords Data Hiding, IWT, PSNR, QR Code, Scrambling

1. Introduction

The amount of multimedia data being transmitted online is growing rapidly due to the swift development of internet technologies [1, 2]. This multimedia data can be of personal, public, governmental or military use. Since the governmental or military data is very sensitive in nature; it faces significant threats while being transmitted online. Thus, information security over the internet is one of the

most researched topics these days to not only protect the privacy of the individuals but also for national security and safety; and researches have taken a lot of data protection measures to transmit the multimedia data securely over the internet. Data security deals with shielding of data from unauthorized and illegal access [3]. These sensitive and critical data must be securely transmitted over the channel. For this purpose, network-specific encryption of data is not trustworthy and consistent; hence, we need an additional method to protect this sensitive data when it is transmitted over the channel so that the secret data cannot be retrieved even when the network routes have been accessed by an unauthorized party.

One of the most basic methods of data security is cryptography and it is one of the most researched topics in the literature [4]. Cryptography is an effective process of storing and transmitting sensitive multimedia data over a public channel or unsecured network where only the targeted individuals can read or process the data. At the receiver side, the cryptic data has to be then decrypted using various algorithms to reveal the original data.

Until recent years, researchers were encrypting texts directly using the basic text cryptosystems. Nowadays, since the size and nature of any multimedia file, be it audio or image is far greater and more complex as compared to that of a text, we can no longer apply these primitive text encryption methods for cryptography.

We can classify image encryption techniques into two broad methods, namely, lossy and lossless [5]. The original encryption methods are quite distorted in the lossy encryption methods; thus, the resulting decrypted image output is slightly different from the original input image. This slight distortion is acceptable considering the characteristics of human perception and the image can be used in a variety of applications [6]. However, the lossless encryption techniques are much more appreciated in applications that involve highly classified images such as in medical imaging; aerospace, satellite imaging, banking and military operations; which require a precise retrieval of

the secret data. Like any other technology, cryptography is not fool proof. Hence, we require several upgradations in the cryptic standards so that the developed cryptography algorithms do not become obvious to the third party while using the recent advancements in computing.

Another form of information hiding method, which uses a digital watermark, embedded by the sender to authenticate the genuineness of the information shared is known as watermarking [7]. Watermarking is used as a platform for security to protect the data from infringements. Digital watermarks are mainly used for copyright protection, authentication, and tampering recovery.

The most secure one-on-one communication technique is steganography [8-11]. The principle behind steganography simple, the unauthorized party must not have a clue of what is being transferred. The art of hiding things is known as steganography, where the secret data is concealed in the cover media while ensuring that the distortion in the cover media is minimum. If the amount of distortion increases, then it may give the third party a hint that some secret data may be stored and be transmitted in the cover media. The cover media can contain images, audio or video to hide secret data.

We can segregate steganography into irreversible and reversible steganography [12]. We use reversible steganography where the original image is sensitive and cannot be modified such as in defence, satellite a and medical imaging. In recent years, steganography based on adaptive pixel selection [1] and other methods like EMD is popular as compared to the previously used primitive steganography which was primarily based on difference expansion and histogram shifting.

We can expect little to moderate distortion in irreversible steganography. The cover media, which are used to carry the secret message, are known as the stego media. It uses the perceptibility of the human eye and transmits the stego media over the channel after intelligently embedding it. At the receiver end, we can easily retrieve the data using various algorithms. Steganography can be classified as spatial domain steganography and transform domain steganography. Both have many data embedding and compatible extraction techniques. Transform-domain steganography is more robust and prone to steganalysis while spatial-domain steganography deals with high data-holding capacity.

The most common steganographic technique in the spatial domain is the least significant bit (LSB) substitution [13, 22]. The LSB method is not prone to steganalysis tools and can be easily detected, despite having an appreciable embedding capacity. Thus, various work has been going on to make LSB prone to steganalysis. LSB Matching Revisited [14] (LSBMR) is one of the techniques being researched upon to solve this problem along with a series of methods which include the combination of Pixel Value Differencing (PVD) and LSB to improve the steganography performance [15].

Discrete Fourier Transform (DFT), Discrete Cosine

Transform (DCT) and Discrete Wavelet Transform (DWT) [16] are the most commonly used signal transforms. DWT is more preferred in steganography due to its spatial frequency local characteristics and multiresolution analysis compared to DCT [17, 18].

Transfer domain stenography involves hiding of data in transfer domain that is usually in Discrete Cosine Transform (DCT), Fourier Transform, Fast Fourier Transform or Discrete Wavelet Transform (DWT), which is different from spatial domain stenography. For instance, Discrete Wavelet Transform of image [19,20] can be shown in terms of its coefficients in the frequency domain. It represents the repetitive nature of the image pixels. The process of hiding the secret image on the transformed image using DWT is called Transform domain stenography [21]. The image is brought back in the spatial domain resulting in stego image. This process is called Inverse DWT. The user follows the same algorithm to retrieve back the original image. Increased security and better data obscurity are some advantages of transform domain stenography. This error can be reduced further by implementing some optimization techniques.

Imperceptibility is the factor that defines the complication involved to find the dissimilarity between stego and cover image [4]. A possible alteration in the output image with the secret image hidden can be estimated by PSNR- Peak Signal to Noise Ratio. Performance of stenography algorithms, security and their capacity can be assessed by multiple indicators. Embedding capacity refers to the amount of secret message that can be embedded in an image.

The remaining paper is segregated as follows. Section 2 deals with the previous works that have been done in steganography with LSB encryption, graph theory, and Haar IWT. Section 3 deals with the proposed algorithm. Section 4 and 5 contain the observed results, discussion and conclusions respectively.

2. Related Work

2.1. Least Significant Bit Substitution

LSB substitution means taking a cover image and substituting its least significant bits with the most-significant bits of the secret image. Over the past few years, data hiding using LSB substitution has gained tremendous popularity due to its large data holding capacity. LSB substitution along with post-processing of the stego images like optimum pixel adjustment process can give us high-quality stego images with a good PSNR and low computational complexity.

The process of hiding data in cover images using the simple least significant bit substitution method is explained below.

Let the cover image I of $M_I X N_I$ pixels be the 8-bit

grayscale image which can be represented as:

$$I = \{x_{ij} | 0 \le i < M_I, \quad 0 \le j < N_I, \quad x_{ij} \in \{0,1,2,...,255\}\}$$
 (1)

The secret message to be embedded consisting of n-bits can be given as

$$M = \{m_i | 0 \le i < n, \ m_i \in \{0,1\}\}$$
 (2)

Now, if the n-bit secret message M has to be embedded in the k-rightmost bits of the cover image I, the secret message M has to be rearranged to form a k-bit virtual image M'represented as:

$$M' = \{m'_i | 0 \le i < n', m'_i \in \{0, 1, ..., 2^k - 1\}\}$$
 (3)

where $n' < M_I * N_I$. The mapping between the embedded message $M' = \{m'_i\}$ and the n-bit secret message $M = \{m_i\}$ is given as follows:

$$m'_{i} = \sum_{j=0}^{k-1} m_{i*k+j} X 2^{k-1-j}$$
 (4)

Bit plane slicing of the cover image is done to separate all the bit planes of the image. Then, the least significant bit plane, i.e., the 0-bit plane of the cover image is replaced with the secret message m_i . The cover image is then reconstructed to form the stego image as:

$$x'_{l_i} = x_{l_i} - x_{l_i} \mod 2^k + m'_i$$
 (5)

where x'_{l_i} is the modified stego-pixel.

At the extraction side, the k-bit LSB of the selected pixels can be readily extracted and recovered by using the following equation:

$$m'_i = x_{l_i} \mod 2^k \tag{6}$$

2.2. Graph Theory and Graceful Graph Generation

The study of points and lines is known as graph theory. A graph G (V, E) can be represented as a function of vertices (V) which are linked with lines or curves called edges (E). In steganography, nodes can be considered as the pixels or pixels coefficients and the connection between two nodes are called edges. Hence, using graph theory, we can generate random traversing algorithms which can be in turn used in steganography algorithms as shown by V. Thanikaiselvan, et. al.

2.3. Method for Graph Generation

As discussed above, the pixel coefficients of an image are known as nodes, and we can easily derive four series S_1 , S_2 , S_3 , S_4 as mentioned below using the number of Nodes (N):

$$S_1 = [2m, 2m - 1, 2m - 2, ..., m + 1]$$
 (7)

where 'm' is an integer and can be derived by m=N/4 where N is the number of nodes in the image. N must be a multiple of 4.

$$S_2 = [m_s(1) + \rho(1), m_s(2) + \rho(2), ..., m_s(m) + \rho(m)]$$
 (8)

where $m_s(i)$ is the i^{th} member of the series $\{m, m-1, m-2, ..., (m-N/4)+1\}$ and $\rho(i)$ is the i^{th} member of the series (ρ) which can be found by the random assembling of the constituents of " m_s ".

 S_3 can be generated with the elements of " ρ " as:

$$S_3 = [\rho(m), \rho(m-1), ..., \rho(1)]$$
 (9)

 S_4 is a series of "m" zeroes:

$$S_4 = [0,0,0,...]$$
 (10)

We can generate a graceful graph with the help of Node Sequence (N_S), Edge Numbers (E), series (S) and a new series 'A' which can be calculated by adding 'E' and 'S'.

2.4. Illustration of Graph Generation

An illustration of graceful graph generation is given as below:

Suppose a 4*4 matrix with sixteen elements considered. Thus, N=16.

Step 1. Generating series S_1 . Since N is given, m can be calculated by m=N/4. Therefore, m=4. Thus, S_1 is given by:

$$S_1 = [8,7,6,5] \tag{11}$$

Step 2. Using the above equation, m(s) = [4,3,2,1], and ρ can be obtained by the random arrangement of m(s), $\rho = [2,4,3,1]$. Thus, S_2 can be generated as:

$$S_2 = [6,7,5,2]$$
 (12)

Step 3. Similarly, S_3 can be generated by substituting ρ in the above equation.

$$S_3 = [1,3,4,2] \tag{13}$$

Step 4. Since q=4, S_4 can be generated as:

$$S_4 = [0,0,0,0] \tag{14}$$

Step 5. Therefore, S can be generated by concatenating $[S_1, S_2, S_3, S_4]$

$$S = [8765675213420000]$$
 (15)

Step 6. Generation of graceful graph table for random coefficient selection is given below:

Step 6(a). Calculating Node Sequence (N_S) depending on the total nodes (N) and tabulating it:

$$N_S = [1, 2, 3, 4, ..., N]$$
 (16)

Step 6(b). Formulating the edge series E= [1, 2, 3, ..., N-1] where the first member of E will be placed directly below the second member of N_S .

Step 6(c). Arranging S in a table with the first element of S under the first element of N_S and then, striking out the first element of S.

Step 6(d). Calculating A = E + S

Step 6(e). A graceful graph is obtained using series S and A

Table 1 shows the generated graceful graph table.

Node Sequence (N_S)	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
Edge Number (E)		1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
S	8	7	6	5	6	7	5	2	1	3	4	2	0	0	0	0
A=E+S		8	8	8	10	12	11	9	9	12	14	13	12	13	14	15

Table 1. Generation of graceful graph table

The graceful graph generated from table 1 is shown in figure 1.

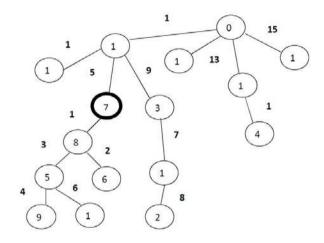


Figure 1. Graceful Graph

The graceful graph matrix R' that can be derived from the graceful graph is shown below:

$$R' = \begin{bmatrix} 7 & 8 & 6 & 5 \\ 9 & 12 & 11 & 10 \\ 3 & 2 & 14 & 4 \\ 1 & 16 & 13 & 15 \end{bmatrix}$$

2.5. Haar Integer Wavelet Transform

Haar IWT can be used to derive the coefficients of the cover image and then the least significant bit substitution can be applied in a lossless manner to get the stego image. After the one-dimensional Haar decomposition of the image, we get the following four parts namely, Approximation (LL), Horizontal (LH), Vertical (HL) and diagonal (HH). The decomposition of the images in the Haar transform is described in figure 2.

During the decomposition of an image with M rows and N columns, using wavelet transform, we assume that every row of size N is a one-dimensional signal. After applying Haar transform, the one-dimensional signal splits into two equal fragments of size $\frac{N}{2}$. This process is known as row processing and is repeated for the remaining M rows to form two signal bands, namely, the high-frequency band (H) and the low-frequency band (L) of size $M * \frac{N}{2}$ using the following equations:

$$H = (C_o - C_e) \tag{17}$$

$$L = \left(C_e + Floor\left(\frac{H}{2}\right)\right) \tag{18}$$

After the row processing, we can separate the odd and even rows of L and H bands as:

$$H_{odd} = Odd \ rows \ of \ H$$

$$L_{odd} = Odd \ rows \ of \ L$$

$$H_{even} = Even \ rows \ of \ H$$

$$L_{even} = Even \ rows \ of \ L$$

$$(19)$$

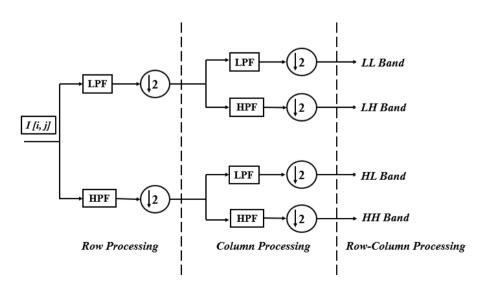


Figure 2. Forward Haar IWT Decomposition of an Image

The H and L bands are then reapplied to forward Haar transform independently, assuming that every column of size M as a one-dimensional signal. This process is known as column processing and it further segregates the H and L sub-bands into the abovementioned LL, LH, HL and HH bands, and each of size $\frac{M}{2} * \frac{N}{2}$ uses the following equations and is described in figure 3:

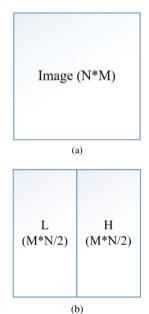
$$LH = L_{odd} - L_{even}$$

$$LL = L_{even} + Floor\left(\frac{LH}{2}\right)$$

$$HH = H_{odd} - H_{even}$$

$$HL = H_{even} + Floor\left(\frac{HH}{2}\right)$$
(20)

In the decomposed cover image, after application of Haar transform; the LL sub-band contains the most important and substantial details when compared to the original image.



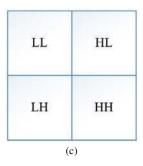


Figure 3. (a) Original Input Image (b) Image after forward IWT row processing (c) Image after forward IWT column processing

3. Proposed Work

Inspired by the integer wavelet transform and graph theory, the proposed method is subdivided into three parts which use different sets of algorithms for data hiding. Each algorithm, in turn, uses a set of keys which have to be used at the extraction site as well to retrieve the desired outputs. The embedding and extraction procedures of the proposed method are discussed here within.

3.1. Embedding Procedure

The entire embedding procedure can be described in figure 4. The multimedia file to be hidden is first converted to a QR-like image by first converting the file to binary and then finally converted to a QR (Quick Response)-like image.

Figure 5 shows the QR-like image of a sample text message. This QR-like image generated is now considered the secret multimedia file.

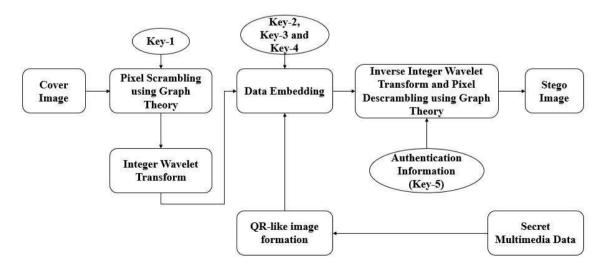


Figure 4. Block Diagram of the Embedding Procedure

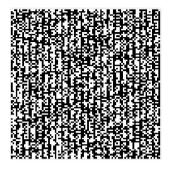


Figure 5. QR-Like Image

3.2. Pixel Scrambling Using Graph Theory

Any image can be denoted as an n-dimensional matrix. The cover image is scrambled using the graceful graph matrix as mentioned above, where every node of the graph is considered as a pixel of the cover image and each node is mapped in the image matrix using the graceful graph formed. To illustrate this, let us consider R to be a 4*4 matrix in which all the matrix elements are considered as image pixels (or) nodes of the graph.

$$R = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 9 & 10 & 11 & 12 \\ 13 & 14 & 15 & 16 \end{bmatrix}$$

Using graph theory, each pixel (or) node of the matrix can be mapped to new locations as shown in R'. This gives us a scrambled image at the output as shown. The mapping done using graph theory can be described in figure 6.

$$R' = \begin{bmatrix} 7 & 8 & 6 & 5 \\ 9 & 12 & 11 & 10 \\ 3 & 2 & 14 & 4 \\ 1 & 16 & 13 & 15 \end{bmatrix}$$

1				
X	7	8	6	5
	9	12	11	10
***	(3)	(2)	4	14
	Y(1)	(16)	13	(15)
			×>	7

Figure 6. Mapping of pixel coefficients of an image using graph theory

Since graph theory leads to very less computational complexity, we can apply the same algorithm to the output of the previous operation to get extensive scrambling of the images. The degree of scrambling depends on the user as they can scramble the input image for 'n' times. Thus, a key can be generated and sent to the extraction site as the degree of scrambling.

3.3. Integer Wavelet Transform

The one-dimensional Haar Integer Wavelet Transform is applied on the scrambled image output of the graph theory. After applying both the row and column processing of the forward Haar transform, the input image (I) is decomposed into four subsequent sub-bands LL, LH, HL, and HH. After the forward Haar transform, of the input image, data can be embedded in the less significant sub-bands of the decomposed image, namely, LH, HL, and HH as shown in figure 7.

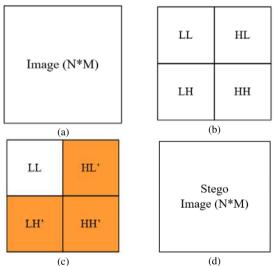


Figure 7. (a) Input Image (b) Image after forward IWT decomposition (c) Image after Data Embedding (d) Stego image after inverse IWT

The LL band is not disturbed while embedding data because it contains sensitive information regarding the cover image. The extent of data embedding in the HL, LH and HH sub-bands decides the PSNR of the input cover image and the generated stego image. An example of the Haar transform decomposition is shown in figure 8.

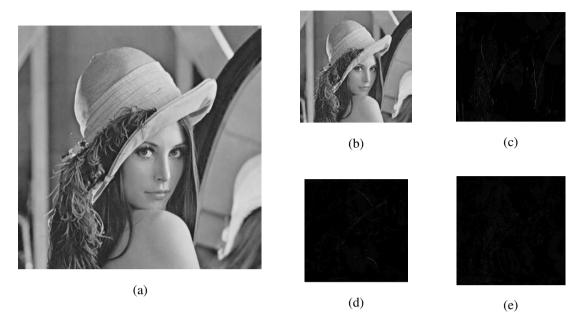


Figure 8. (a) Cover Image: Lena (b) LL (c) HL (d) LH (e) HH forward Haar decomposition of (a)

The stego image is generated after inverse column processing and inverse row processing of the image in which data is embedded.

3.4. LSB Data Embedding

Least Significant Bit substitution is used to embed the data in the cover image after forward IWT decomposition of the cover image. Here, the bit planes of the secret image are separated along with the HL, LH and HH decomposed planes of the cover image.

The bit plane slicing of the cover image is done by:

$$C_i = Mod\left(floor\left(\frac{c}{2^i}\right), 2\right)$$
 (21)

where C is the cover image, C_i is the i^{th} bit plane of C. We can separate the most significant bits of the message to be

hidden in the same procedure. The MSB of the secret message image (I) can be embedded in the cover image by:

$$C'_{l_i} = C_{l_i} - C_{l_i} mod \ 2^k + m'_i$$
 (22)

Here, the user can define the number of bits which have to be embedded in each plane of the image. It is not recommended to embed more than 4 bits in each plane for an optimal quality stego image otherwise, the image quality degrades.

3.5. Extraction Procedure

The sequence of the secret data extraction method closely resonates with the embedding method. The following block diagram in figure 9 shows the procedure of how to retrieve the QR-like image and in turn, the secret message from the stego image is formed.

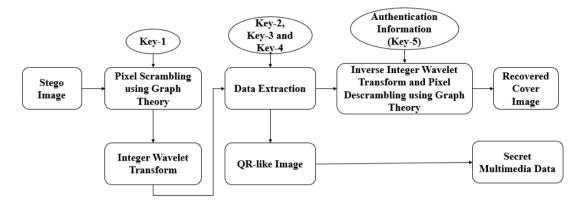


Figure 9. Block Diagram of the Extraction Procedure

The stego image is again scrambled with the same degree as used during the embedding procedure. After the pixel scrambling, the stego image is again decomposed in LL, LH, HL and HH sub-bands. Then, the embedded secret message bits can be recovered from the LH, HL and HH bands by using the following equation:

$$m'_{i} = C_{i} \bmod 2^{k} \tag{23}$$

where m'_i represents the extracted message bits of the secret image. Inverse IWT can be applied to the remaining image to recover the cover image again.

4. Results

The proposed method has been tested on ten different grayscale images, each of size 512*512. The performance characteristics of the output stego image have been compared with the cover image using Peak Signal to Noise Ratio (PSNR), Mean Square Error (MSE), Normalized Cross-Correlation (NCC) and Structural Similarity Index Matrix (SSIM) which are shown below from equations (24-27) where both the sizes of the cover image (C) and the stego image (S) are M*N.

$$MSE = \frac{1}{MN} \sum_{i=1}^{M} \sum_{j=1}^{N} \left(C_{i,j} - S_{i,j} \right)^{2}$$
 (24)

$$PSNR = 10log_{10} \frac{255*255}{MSE} dB$$
 (25)

$$NCC = \frac{\sum_{i=1}^{M} \sum_{j=1}^{N} C(i,j) * S(i,j)}{\sum_{i=1}^{M} \sum_{j=1}^{N} C(i,j)^{2}}$$
(26)

$$SSIM(C,S) = \frac{(2\mu_C \mu_S + x_1) * (2\sigma_{CS} + x_2)}{(\mu_C^2 + \mu_S^2 + x_1) * (\sigma_C^2 + \sigma_S^2 + x_2)}$$
(27)

where

 μ_C is the mean pixel value of C;

 μ_S is the mean pixel value of S;

 σ_C^2 is the variance of C;

 σ_S^2 is the variance of S;

 σ_{CS} is the covariance of C and S;

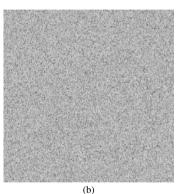
 $x_1 = (k_1 L)^2$ and $x_2 = (k_2 L)^2$ are used to stabilize the division due to the weak denominator;

L is the dynamic range of pixel values;

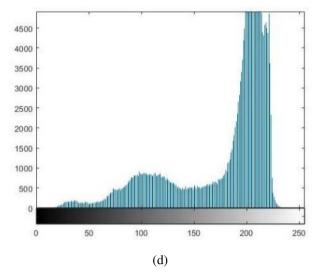
$$k_1 = 0.01$$
 and $k_2 = 0.03$

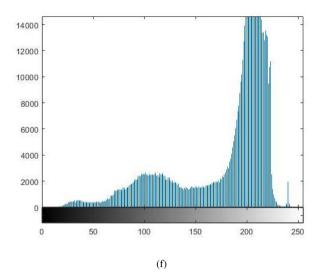
Histogram analysis of the cover image, scrambled image and stego image and their comparison gives us very similar results as observed from figure 10.

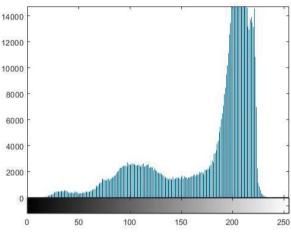












(e)

Figure 10. (a) Original Airplane; (b) Scrambled Airplane; (c) Stego Airplane; (d) Original Airplane-Histogram; (e) Scrambled Airplane-Histogram; (f) Stego Airplane-Histogram

Table 2 compares the MSE, PSNR, NCC and SSIM for the generated output images from the above method with up to 4-bit embedding in the image sub-bands. The average MSE and PSNR of 1-bit embedding (196608 bits per image) are 0.5 and 50 dB respectively. Similarly, for 2-bit embedding (393216 bits per image), 3-bit embedding (589824 bits per image) and 4-bit embedding (786432 bits per image), the average PSNR values observed are 44 dB, 38 dB and 33 dB respectively. Normalized Cross Correlation and Structural Similarity Index Matrix compare the degree of accuracy between the cover and stego image as shown in table 2 which are measured between 0 and 1.

Table 2. Comparison between different images and their observed results

Image Dataset	Bits embedded per sub-band	MSE	PSNR	NCC	SSIM
Airplane	1	0.5629	50.3412	0.9998	0.9868
Airplane	2	2.4227	44.8211	0.9998	0.9211
Airplane	3	11.6128	37.5154	0.9998	0.7893
Airplane	4	42.0913	31.2343	0.9998	0.5864
Barb	1	0.5051	50.1060	0.9996	0.9825
Barb	2	2.1825	44.2831	0.9996	0.9419
Barb	3	10.4805	37.4977	0.9996	0.8341
Barb	4	48.1888	31.3353	0.9996	0.6559
Boat	1	0.6189	49.9454	0.9997	0.9629
Boat	2	2.6783	43.8863	0.9997	0.8892
Boat	3	10.4721	37.9644	0.9997	0.7467
Boat	4	39.4013	32.2097	0.9997	0.5589
Cameraman	1	0.5537	50.7320	0.9996	0.9900
Cameraman	2	2.1708	44.7986	0.9996	0.9633
Cameraman	3	10.0089	38.1609	0.9996	0.8716
Cameraman	4	41.4342	31.9912	0.9996	0.6653
Goldhill	1	0.5521	50.7449	0.9996	0.9876
Goldhill	2	2.1729	44.7944	0.9996	0.9598
Goldhill	3	9.9782	38.1743	0.9996	0.8818
Goldhill	4	41.0669	32.0299	0.9996	0.7092
Ноте	1	0.5363	50.8710	0.9998	0.9342
Ноте	2	2.1951	44.7503	0.9998	0.8261
Ноте	3	10.5031	37.9516	0.9998	0.6302
Ноте	4	35.8253	32.6229	0.9998	0.4154
Lena	1	0.5542	50.7283	0.9996	0.9786
Lena	2	2.1806	44.7791	0.9996	0.9231
Lena	3	9.9233	38.1982	0.9996	0.7771
Lena	4	39.6343	32.1841	0.9996	0.5319
Mandrill	1	0.5436	50.5584	0.9996	0.9887
Mandrill	2	2.1751	44.2978	0.9996	0.9321
Mandrill	3	10.9482	38.2901	0.9996	0.8274
Mandrill	4	38.3695	32.7018	0.9996	0.5880
Peppers	1	0.5132	50.9083	0.9996	0.9814
Peppers	2	2.7942	43.6943	0.9996	0.9246
Peppers	3	10.6401	37.4283	0.9996	0.7994
Peppers	4	39.1964	31.9897	0.9996	0.5772
Tiffany	1	0.6419	49.7352	0.9998	0.9555
Tiffany	2	2.1498	44.1376	0.9998	0.9128
Tiffany	3	11.2976	38.0731	0.9998	0.8432
Tiffany	4	39.4232	31.8128	0.9998	0.5655

The proposed method also leads to minimal computational complexity. The average time required for the entire embedding and extraction combined is found to be 4.5512 seconds. The time complexity of the same can be given in Big-O notation as O(N) since the algorithm is operated sequentially for each pixel.

5. Conclusions

We can prove the high capacity, imperceptibility and robustness of the output stego image form the proposed method in terms of PSNR, MSE, NCC and SSIM which gives us very convincing results as compared to other methods. High capacity embedding is achieved through 4-bit embedding in the image sub-bands. High robustness of the stego image is attained through QR-like image formation of the data to be hidden in the image using graph theory and Haar transform. The proposed method can be adapted with various other data embedding algorithms and their performance can be analyzed as a future extension.

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