



Steganography using Bit Plane Complexity Segmentation and Artificial Neural Network

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Abstract: Steganography is a technique to hide information in some other vessel data without leaving any apparent evidence of vessel data alteration. There are different techniques for steganography namely substitution technique, transform domain technique, spread spectrum technique, statistical method, distortion technique and cover generation technique. The three characteristics which are taken care of are:- capacity, undetectability and imperceptibility. Multilayer neural network is used to achieve image compression. The input pixels are used as target values and the hidden layer output is the compressed image. Our proposed technique uses an image as the vessel data and embeds a secret image that is compressed by artificial neural network in the bit-planes of the vessel image without deteriorating image quality. This technique makes use of the characteristics of the human vision system whereby a human cannot perceive any change in the information in a very complicated binary pattern. Its information hiding capacity can be made large.

Keywords: Steganography, Information hiding, Multilayer neural network, image compression, Hidden layer, Vessel image.

I. INTRODUCTION

To conceal the very existence of transmission we use steganography. Encryption clearly marks a message as containing “interesting” information and the encrypted message becomes subject to attack. So, a typical steganographic technique may use image data as the container of secret information. Some of the techniques use least significant bits of the image pixels to hide the data while others embed the secret information in a specific band of spatial frequency component of carrier. [3] Image compression has been an active area of research since the inception of digital image processing. Compression is achieved by exploiting the redundancy in an image. Artificial Neural Network have found increasing applications in this field due to their noise suppression and learning capabilities. [1]

In this paper we have proposed a technique for Steganography using bit plane complexity segmentation and artificial neural network.

Some of the concepts used are as follows:-

1. Back Propagation Neural Network

An Artificial Neural Network (ANN) is a computational model based on the structure and functions of biological neural networks. They are considered nonlinear statistical data modelling tools where the complex relationships between inputs and outputs are found. [6]

Back Propagation Neural Network is designed and trained using different learning algorithms. The neural network structure shown in Figure 1 has three layers: one input layer, one output layer and one hidden layer. Both of input layer and output layer are fully connected to hidden layer. [2]

The connection weights in the network are gradually adjusted, working backwards from the output layer, to the input layer through the hidden layer, until correct output is produced. The general parameters deciding the performance of back propagation neural network algorithm includes

mode of learning, target values, goal, epochs, learning rate and momentum factors. [2]

Compression is achieved by designing the value of the number of neurons at the hidden layer, less than that of neurons at both input and output layers. The above neural network could be either linear or nonlinear network according to their activation function. [2]

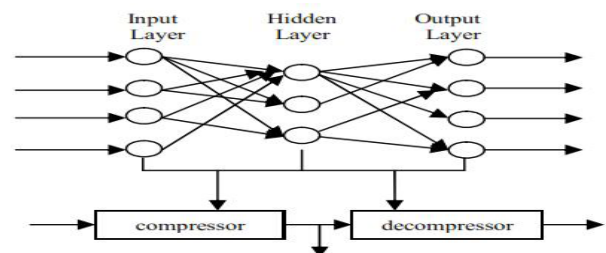


Figure 1

1.1 Network Training Algorithms

Training the network is an important step to get the optimal values of weights and biases after being initialized randomly or with certain values. For training the network, the 128x128 pixels image has been employed. [1]

During training, the weights and biases of the network are iteratively adjusted to minimize the network performance function which is the mean square error for the feed forward networks. The mean square error is calculated as the average squared error between the inputs and targets. [1]

In the basic back propagation training algorithm, the weights are moved in the direction of the negative gradient, which is the direction in which the performance function decreases most rapidly. It can be written as: $X_{k+1} = X_k - \alpha_k g_k + 1$, where X_{k+1} is a vector of current weights and biases, g_k is the current gradient, and α_k is the learning rate. [1]

Training could be started by making target matrix equal to input matrix. Different training algorithms can be used, mainly classified as fast algorithms (example trainbfg,

trainoss, trainlm, etc.) and slow algorithms (example traingd, traingda, etc.)

1.2 Simulation of Results

After training, we obtain a simulated network by the input matrix and the target matrix. In the simulation process two outputs are obtained, the hidden layer output and the output layer output. The hidden layer produces a matrix of 16x256, whereas the output layer produces a matrix of 64x256.

Hidden layer output is the compressed image where $h_i < n_i$, where h_i is the number of neurons in hidden layer and n_i is the number of neurons in the input layer. [2]

In the hidden layer matrix, each column should be reshaped into 4x4 pixel blocks, while in the output layer matrix, each column should be reshaped into 8x8 pixel blocks to display both matrices as images. [1]

II. STEGANOGRAPHY

Steganography literally means covered writing and is the art of hiding secret messages within another seemingly innocuous message or carrier. The advantage of steganography is that the intended secret message does not attract attention to itself as an object of scrutiny.

The carrier used is a grey scale image and the hidden image appear to be hidden (or be part of) the *cover image*. Steganography includes the concealment of information within computer files. In digital steganography, electronic communications may include steganographic coding inside of a transport layer, such as a document file, image file, program or protocol. [4]

2.1 BPCS Technique for Steganography:

- Canonical Grey Code

In PBC the substantial portions of the regions on the higher bit planes are relatively flat in color (mostly all 0s or all 1s). This is because of the "Hamming Cliffs" which occur with PBC wherein a small change in color affects many bits of the color value.

In binary:

$$127 = 01111111 \text{ \& } 128 = 10000000$$

In gray code:

$$127 = 01000000 \text{ \& } 128 = 11000000$$

In gray code the representation of adjacent gray levels will differ only in one bit (unlike binary format where all the bits can change). CGC images do not suffer from such Hamming Cliffs.

- Complexity of an image

The complexity of an image (α) is defined by the following: $\alpha = k / \text{The maximum possible B-W changes in the image}$, where, k is the total length of black-and-white border in the image.

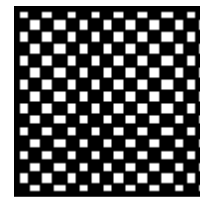
So, value of α ranges over $0 \leq \alpha \leq 1$.

- Conjugation of a binary image:-

It is applied when the complexity of the vessel image block is less than the threshold and cannot be used for embedding the secret data.

$$P^* = P \text{ XOR } W_c$$

where P^* is a conjugated 8x8 block of vessel image, P is an informative (i.e. less complex) 8x8 block of vessel image, W_c is a complex binary pattern with a white pixel at the upper left corner position.



W_c pattern

After conjugation, complexity of the vessel image block increases and thus can be used for embedding secret data.

$$\alpha(P^*) = 1 - \alpha(P). \quad [3]$$

III. PROPOSED TECHNIQUE

With this proposed technique, we can hide a larger size secret image into vessel image after compressing it using artificial neural network. This is extremely important for efficient storage and transmission of image data. Multilayer neural network has many advantages like noise suppression, learning capabilities and can accurately approximate to any linear or non-linear function. It is thus perfect to be used for image compression.

In [1] the Image compression technique and in [2] BPCS technique is described. The proposed technique is described in algorithm 1 and algorithm 2.

Algorithm 1 for the Image Compression:-

1. Take a 128x128 grey scale secret image. Divide it into 8x8 pixel blocks and reshape each one into 64x1 column vector.
2. Arrange the column vectors into a matrix of 64x256, which will be used as input matrix.
3. Let the target matrix equal to the input matrix obtained in step 2.
4. Choose a suitable learning algorithm such as `traincgf` and parameters like `goal - 0.001` to start training the neural network using neural network toolbox of matlab.
5. Train and simulate the network with the given input matrix and the target matrix.
6. Obtain the output matrices of the hidden layer and the output layer.
7. Post-process them to obtain the compressed image and the reconstructed image respectively.

Explanation for the above algorithm is as follows:-

We take a 128x128 grey scale secret image and divide it into 8x8 pixel blocks. Then reshape each one into 64x1 column vector. We arrange these column vectors into a matrix of 64x256, which will be used as input matrix named as sample. Let the target matrix equal to the sample matrix. Then we use the neural network toolbox of matlab to create, train and simulate an artificial neural network.

To create a neural network, the network type used is feed forward back propagation neural network, training function is traincgf, performance function is mean square error and transfer function is purelin. After it is created, we train and simulate the network to obtain the output matrices of the hidden layer and the output layer. The hidden layer produces

a matrix of 16x256, whereas the output layer produces a matrix of 64x256. We post-process them to obtain the compressed image of size 64x64 and the reconstructed image of size 128x128.

Algorithm 2 for the BPCS technique:-

1. Transform the grey scale vessel image from PBC to CGC system.
2. Divide our vessel image into a series of bit planes (1-8).
3. Segment each bit-plane of the vessel image into informative and noise-like regions by using a threshold value ($\alpha=0.3$).
4. Divide the secret file into a series of 8 x 8 secret blocks.
5. If a block (P) of cover image is less complex than the threshold (α), then conjugate it to make it a more complex block (P*). (Provided it is not used for embedding the watermark or the signature)
6. If the block is conjugated, then record this fact in a “conjugation map”. Embed the conjugation map in the noise block with the highest entropy.
7. Embed each secret block into the noise-like regions of the bit-planes (or replace all the noise-like regions with a series of secret blocks).
8. Generate a new stego image using modified bit planes.
9. Convert the stego image from CGC back to PBC.

Explanation for the above algorithm is as follows:-

We take a 256x256 grey scale image which is used as the vessel image. It is converted from PBC to CGC form before starting the algorithm. We divide the vessel image into 8 bit planes using bitget() function of matlab. Then each bit plane is segmented into informative and noise like regions by using a threshold value α . We divide compressed secret file of 64x64 size into 8x8 blocks.

If a block (S) of the vessel image is less complex than the threshold (α_0), then we make it more complex (S*) $\alpha(P^*) = 1 - \alpha(P)$ by **conjugation**. If the block is conjugated, then it is recorded in a “**conjugation map**”. We embed the conjugation map in the noise block of the vessel image with the highest entropy. Embed each secret block into the noise-like regions of the bit-planes. At the end, we convert the stego image from **CGC** back to **PBC**.

IV. EXPERIMENTAL RESULTS



Fig 1.1 a), b) represent original vessel image and secret data respectively.

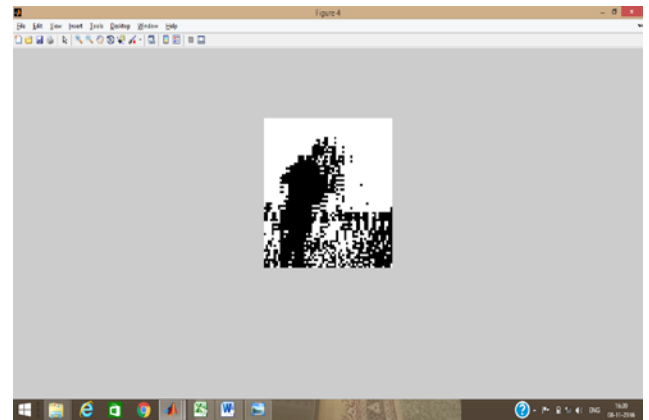


Fig 1.2 shows the compressed secret image.

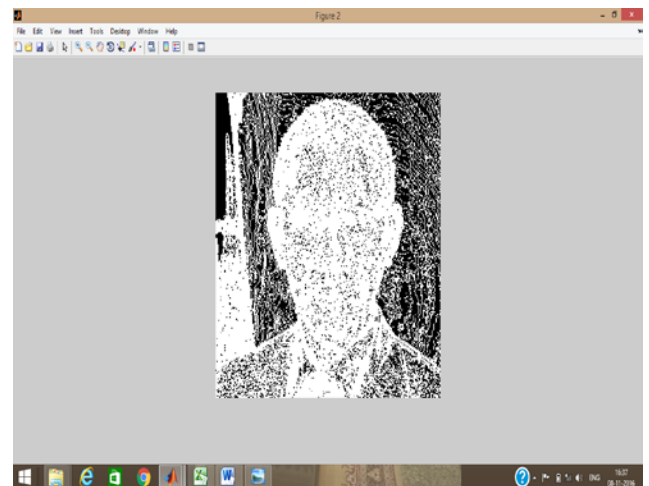


Fig 1.3 shows vessel image in CGC system.

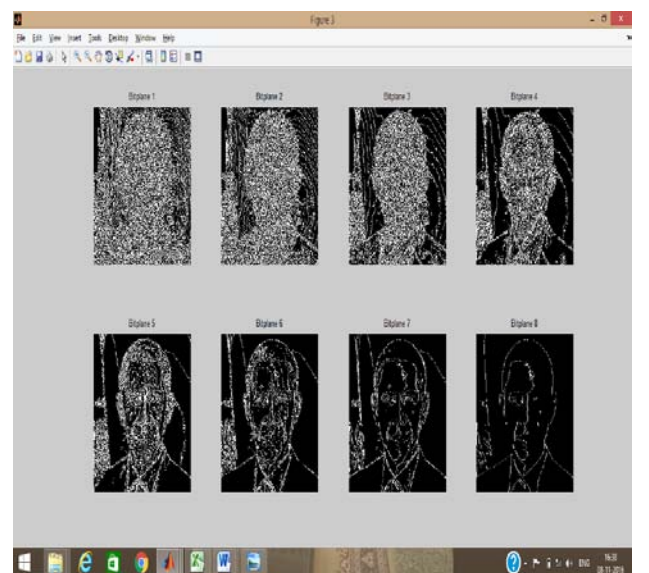


Fig 1.4 shows the bit plane slicing of the vessel image.

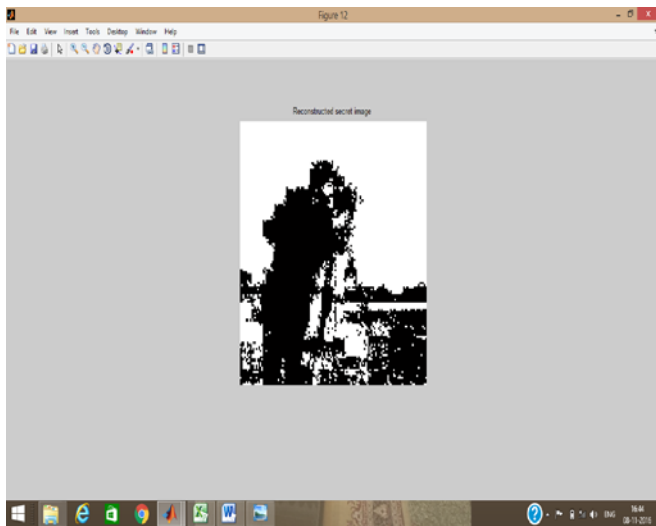


Fig 1.5 shows the image reconstructed from the compressed secret image.

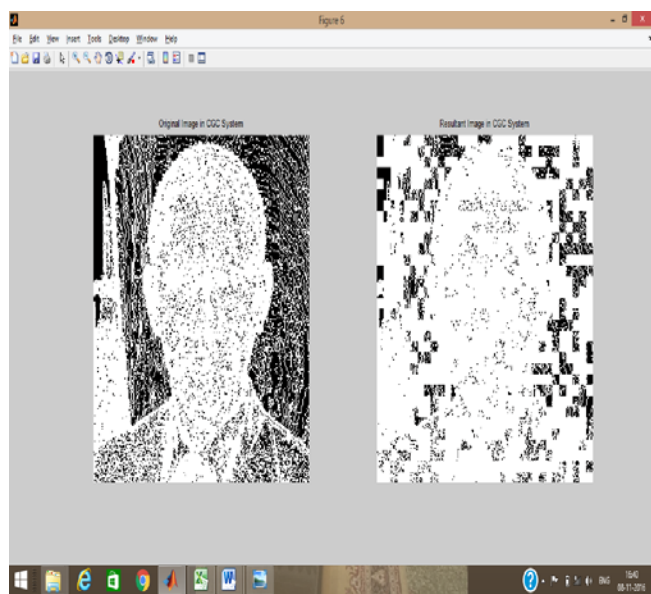


Fig 1.6 a), b) shows the original vessel image and vessel image obtained after embedding secret file in CGC system.



Fig 1.7 a), b) shows the Original vessel image and vessel image which is obtained after embedding the secret file.

V. ANALYSIS

The following table shows PSNR values between original vessel images and vessel images obtained after embedding different secret files in different formats. These PSNR values indicate that the robustness of the technique is preserved.

S.No.	Vessel image used	Secret image used	PSNR Value
1	Test 1.gif	Secret 1.bmp	51.2633
2	Test 2.bmp	Secret 1.bmp	51.2988
3	Test 3.bmp	Secret 1.bmp	51.1832
4	Test 4.jpg	Secret 1.bmp	51.2254
5	Test 5.png	Secret 1.bmp	51.2121
6	Test 6.gif	Secret 1.bmp	51.2966
7	Test 7.jpg	Secret 1.bmp	51.2298
8	Test 8.png	Secret 1.bmp	51.2681
9	Test 9.gif	Secret 1.bmp	51.2617
10	Test 10.jpg	Secret 1.bmp	51.2693
11	Test 1.gif	Secret 2.jpg	51.2461
12	Test 2.bmp	Secret 2.jpg	51.3007
13	Test 3.bmp	Secret 2.jpg	51.2058
14	Test 4.jpg	Secret 2.jpg	51.2430
15	Test 5.png	Secret 2.jpg	51.2227
16	Test 6.gif	Secret 2.jpg	51.2797
17	Test 7.jpg	Secret 2.jpg	51.2352
18	Test 8.png	Secret 2.jpg	51.2769
19	Test 9.gif	Secret 2.jpg	51.2696
20	Test 10.jpg	Secret 2.jpg	51.2514
21	Test 1.gif	Secret 3.png	51.2887
22	Test 2.bmp	Secret 3.png	51.3089
23	Test 3.bmp	Secret 3.png	51.1771
24	Test 4.jpg	Secret 3.png	51.2356
25	Test 5.png	Secret 3.png	51.1956
26	Test 6.gif	Secret 3.png	51.2808
27	Test 7.jpg	Secret 3.png	51.2575
28	Test 8.png	Secret 3.png	51.2965
29	Test 9.gif	Secret 3.png	51.2727
30	Test 10.jpg	Secret 3.png	51.2593



Test 1.gif



Test 2.bmp



Test 3.bmp



Test 9.gif



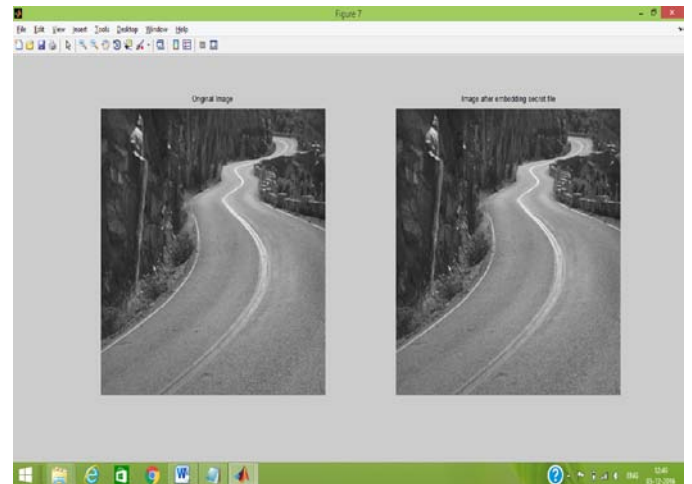
Test 4.jpg



Test 10.jpg



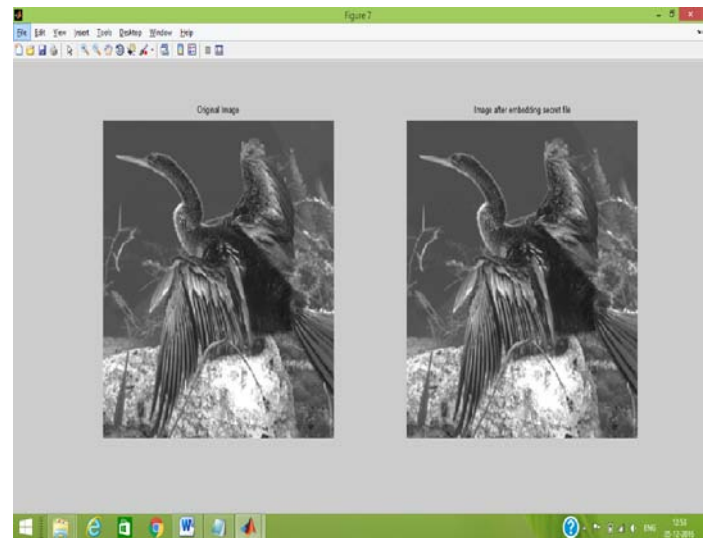
Test 5.jpg



a),b) original image test 8.png and image test 8.png after embedding secret file respectively.



Test 6.gif



a),b) original image test 1.gif and image test 1.gif after embedding secret file respectively.



Test 7.jpg



Test 8.jpg

The following table shows PSNR values between original secret images and reconstructed secret images obtained after training and simulation of neural network. These PSNR values indicate that the robustness of the technique is preserved.

S.No.	Secret Image used	PSNR value
1	Secret 1.bmp	60.2917
2	Secret 2.jpg	60.6900
3	Secret 3.png	60.2443
4	Secret 4.bmp	60.8298
5	Secret 5.png	60.7043
6	Secret 6.gif	59.5828
7	Secret 7.gif	59.1863
8	Secret 8.gif	57.7780
9	Secret 9.gif	59.9990
10	Secret 10.png	58.1925



Secret 1.bmp



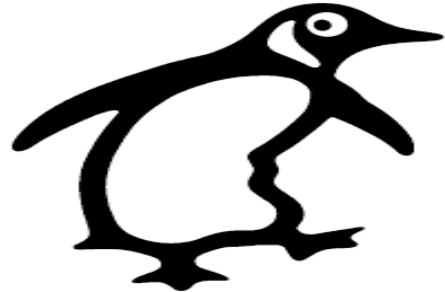
Secret 2.jpg



Secret 3.png



Secret 4.bmp



Secret 5.png



Secret 6.gif



Secret 7.gif



Secret 8.gif



Secret 9.gif



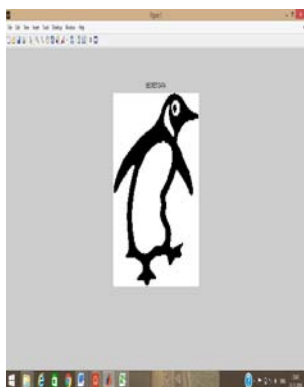
Secret 10.png



a)Original Barbara.png



b)Reconstructed barbara.png



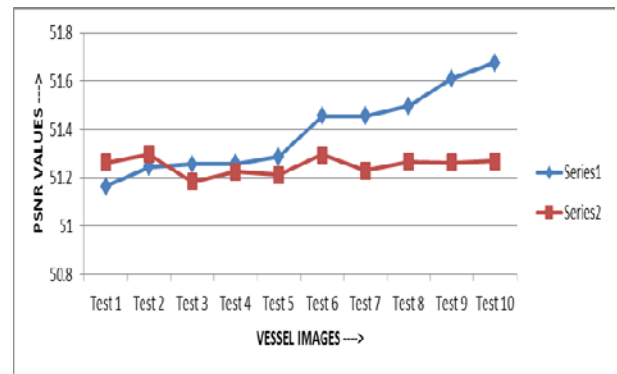
a) Original p1.png



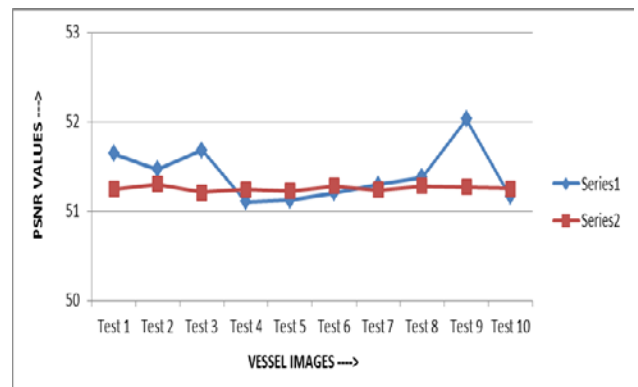
b) Reconstructed p1.png

VI. COMPARISON WITH THE EXISTING WORK

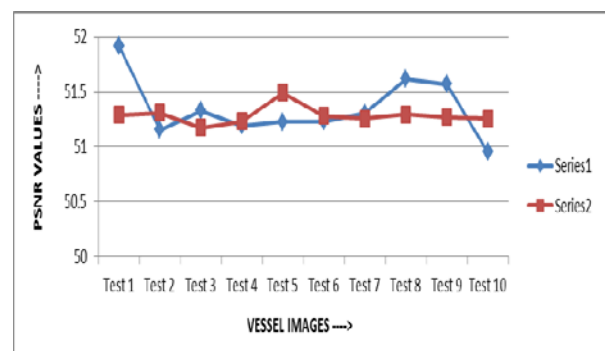
- The following graph indicates the comparison of psnr values of the ten test images (secret image used is secret 1) between the original work and the proposed work.



- The following graph indicates the comparison of psnr values of the ten test images (secret image used is secret 2) between the original work and the proposed work.



- The following graph indicates the comparison of psnr values of the ten test images (secret image used is secret 3) between the original work and the proposed work.



VII. APPLICATIONS

1. TRANSACTION TRACKING

Watermarks record the recipient in each legal sale or distribution of the work. If the work is misused (leaked to the press or illegally distributed), the owner could find out about who is the traitor. Visible watermarking is often adopted in this application but invisible watermark is even better.

2. FORENSIC AND PIRACY DETERRENCE

Forensic application embeds the identity of a recipient into an asset copy at the time it is produced or transmitted. It helps companies create a powerful deterrence from leaking controlled content either

maliciously or unintentionally, quickly and accurately identify the source of leaked content, provide irrefutable evidence of content misuse in support of legal action, gain visibility over where and how their content is being accessed.

3. COPY CONTROL

It combines every content recorder with a watermark detector. When a copy-prohibit watermark is detected, the recording device will refuse to copy.

4. DOCUMENT AND IMAGE SECURITY

Images and documents are distributed to remote offices, agencies, distributors, dealers and must be managed to ensure confidential information is not leaked before the launch date.

VIII. CONCLUSION

- The most important point in this technique is that human eye cannot perceive any change in the information in the bit-planes of an image, thus protecting the copyright of image.
- By using artificial neural network for image compression, we are able to use a larger size secret image for hiding in the vessel image.

Thus this technique is effective for both storage and transmission of data. Also with BPCS steganography, there is an increase in level of secret and secure communication.

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