

Steganography and Image Enhancement- A Proposed Method

Nitika Kapoor¹, Aashima² and Navneet Malik³

^{1,2}Lovely Professional University, Phagwara,

²Rayat & Bahra Institute of Engineering & Bio-Technology, Sahauran
er.nitikakapoor@gmail.com, er.aashima@yahoo.co.in

Abstract: In this paper we have proposed a method to combine the features of image enhancement and Steganography. Various still images will be used on which the tests will be implemented.

Keywords: Image enhancement, Steganography

1. Introduction

Removing and reducing impulse noise is very active research area in image processing. Present day applications require various kinds of images and pictures as sources of information for interpretation and analysis. Whenever an image is converted from one form to another, some form of degradation occurs at the output. The output image has to undergo a process called image enhancement. An effective method for image enhancement was presented by Russo, which was controlled by tuning of one parameter.

Digital communication has become an essential part of infrastructure nowadays, a lot of applications are Internet-based and in some cases it is desired that the communication be made secret. Two techniques are available to achieve this goal: one is cryptography, where the sender uses an encryption key to scramble the message; this scrambled message is transmitted through the insecure public channel, and the reconstruction of the original, unencrypted message is possible only if the receiver has the appropriate decryption key. The second method is steganography, where the secret message is embedded in another message. Using this technology even the fact that a secret is being transmitted has to be secret.

Our method is to combine these two techniques.

2. Proposed Work

Steganography and Enhancement are the two broad categories in the field of image processing. We are tried to combine these two fields. The method is discussed here.

2.1 Noisy Image

The method we are going to develop will be for the noisy image. We assume that the image contain the salt and

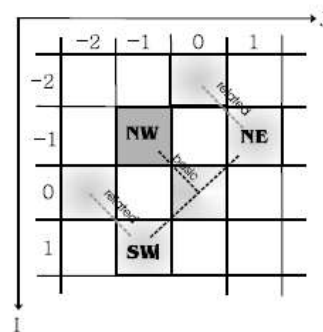
pepper noise.

Removal of Impulse noise

We start from a gray scale image in order to better explain how the new algorithm is constructed. Let the grayscale image be represented by a matrix F of size $N1 \times N2$, $F = \{F(i, j) \in \{0, \dots, 255\}, i = 1, 2, \dots, N1, j = 1, 2, \dots, N2\}$. Our construction starts with the introduction of the similarity function $\mu: [0; \infty) \rightarrow R$. We will need the following assumptions for μ :

- (1) μ is decreasing in $[0; \infty)$,
- (2) μ is convex in $[0; \infty)$,
- (3) $\mu(0) = 1, \mu(\infty) = 0$.

In the construction of filter, the central pixel in the window W is replaced by that one, which maximizes the sum of similarities between all its neighbors. Basic assumption is that a new pixel must be taken from the window W .



For each pixel (i, j) of the image (that isn't a border pixel) use a 3×3 neighborhood window. Each neighbor with respect to (i, j) corresponds to one direction $\{NW = North West, N = North, NE = North East, W = West, E = East, SW = South West, S = South, SE = South East\}$. Each such direction with respect to (i, j) can also be linked to a certain position.

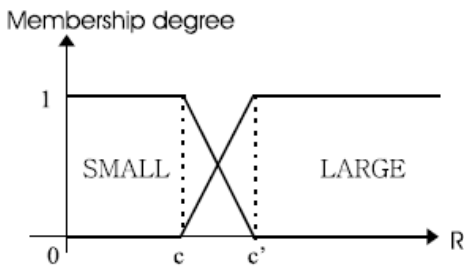
INVOLVED GRADIENT VALUES TO CALCULATE THE FUZZY GRADIENT.

R	basic gradient	related gradients
NW	$\nabla_{NW}A(i,j)$	$\nabla_{NW}A(i+1,j-1), \nabla_{NW}A(i-1,j+1)$
N	$\nabla_NA(i,j)$	$\nabla_NA(i,j-1), \nabla_NA(i,j+1)$
NE	$\nabla_{NE}A(i,j)$	$\nabla_{NE}A(i-1,j-1), \nabla_{NE}A(i+1,j+1)$
E	$\nabla_EA(i,j)$	$\nabla_EA(i-1,j), \nabla_EA(i+1,j)$
SE	$\nabla_{SE}A(i,j)$	$\nabla_{SE}A(i-1,j+1), \nabla_{SE}A(i+1,j-1)$
S	$\nabla_SA(i,j)$	$\nabla_SA(i,j-1), \nabla_SA(i,j+1)$
SW	$\nabla_{SW}A(i,j)$	$\nabla_{SW}A(i-1,j-1), \nabla_{SW}A(i+1,j+1)$
W	$\nabla_WA(i,j)$	$\nabla_WA(i-1,j), \nabla_WA(i+1,j)$

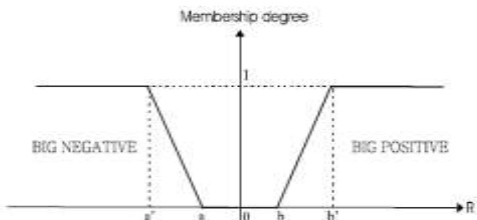
Each direction R corresponds to central position. Column 2 gives the basic gradient for each direction; column 3 gives the two related gradients. The fuzzy gradient value for direction R is calculated by following fuzzy rule:

IF $|\nabla_R A(i,j)|$ is large AND $|\nabla'_R A(i,j)|$ is small
 OR
 $|\nabla_R A(i,j)|$ is large AND $|\nabla''_R A(i,j)|$ is small
 OR
 $\nabla_R A(i,j)$ is big positive AND ($\nabla'_R A(i,j)$
 AND $\nabla''_R A(i,j)$) are big negative
 OR
 $\nabla_R A(i,j)$ is big negative AND ($\nabla'_R A(i,j)$
 AND $\nabla''_R A(i,j)$) are big positive
 THEN $\nabla^F_R A(i,j)$ is large

The membership function used are LARGE (for the fuzzy set large), SMALL (for the fuzzy set small), BIG POSITIVE (for the fuzzy set big positive) and BIG NEGATIVE (for the fuzzy set big negative).



The above graph shows the membership function for fuzzy set SMALL and LARGE.



The above graph shows the membership function for fuzzy set BIG NEGATIVE and BIG POSITIVE.

The pixels of the image are arranged in these membership functions. The noisy pixels are then sort out and form the member of the function more or less impulse noise. The noisy pixel values are then changed according to the

following formula:

$$F(i,j) = \frac{\sum_{h=-1}^1 \sum_{l=-1}^1 [1 - \mu(A(i+h,j+l))] A(i+h,j+l)}{\sum_{h=-1}^1 \sum_{l=-1}^1 1 - \mu(A(i+h,j+l))}$$

Improving contrast of the image

For improving the contrast of the image following steps are done: setting the shape of membership function (regarding to the actual image) setting the value of fuzzifier Beta calculation of membership values modification of the membership values by linguistic hedge generation of new gray-levels.

Using the notation of fuzzy sets, we can write,

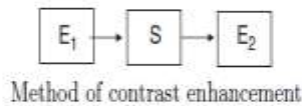
$$X = \begin{pmatrix} \mu_{11}/x_{11} & \mu_{12}/x_{12} & \dots & \mu_{1M}/x_{1M} \\ \mu_{21}/x_{21} & \mu_{22}/x_{22} & \dots & \mu_{2M}/x_{2M} \\ \vdots & \vdots & & \vdots \\ \mu_{N1}/x_{N1} & \mu_{N2}/x_{N2} & \dots & \mu_{NM}/x_{NM} \end{pmatrix}$$

Where $0 \leq \mu_{mn} \leq 1, m = 1, 2 \dots M, n = 1, 2 \dots N$.

Contrast within an image is measure of difference between the gray-levels in an image. The greater the contrast, the greater is the distinction between gray-levels in the image. Images of high contrast have either all black or all white regions; there is very little similar gray-levels in the image, and very few black or white regions. High-contrast images can be thought of as crisp, and low contrast ones as completely fuzzy. Images with good gradation of grays between black and white are usually the best images for purposes of recognition by humans.

The object of contrast enhancement is to process a given image so that the result is more suitable than the original for a specific application in pattern recognition. As with all image-processing techniques we have to be especially careful that the processed image is not distinctly different from the original image, making the identification process worthless. The technique used here makes use of modifications to brightness membership value in stretching or contracting the contrast of an image.

Many contrast enhancement methods work as shown in the figure below, where the procedure involves primary enhancement of the image, denoted with an E1 in the figure, followed by a smoothing algorithm, denoted by an S, and a subsequent final enhancement, step E2.



The function of the smoothing operation of this method is to blur (make fuzzier) the image and this increased blurriness then requires the use of final enhancement step E2. Generally smoothing algorithms distribute a portion of the intensity of one pixel in the image to adjacent pixels. This distribution is greatest for pixels nearest to the pixels being smoothed, and it decreases for pixels farther from the pixel being smoothed.

2.2 Steganography

Steganography embeds a secret message in a cover message, this process is usually parameterized by a stego-key, and the detection or reading of embedded information is possible only having this key.

We will implement steganography on noisy and low contrast images. We have opted two methods for the same.

LSB

LSB is the lowest bit in a series of numbers in binary. e.g. in the binary number: 10110001, the least significant bit is far right 1.

The LSB based Steganography is one of the steganography methods, used to embed the secret data in to the least significant bits of the pixel values in a cover image. e.g. 240 can be hidden in the first eight bytes of three pixels in a 24 bit image.

DCT

DCT coefficients are used for JPEG compression. It separates the image into parts of differing importance. It transforms a signal or image from the spatial domain to the frequency domain. It can separate the image into high, middle and low frequency components.

METHOD

Firstly, we will take the noisy and low contrast images. At the sender end we will embed an image in the said image.

Then, the noise will be removed from the image and the embedded image is extracted.

The stego image is extracted using two algorithms LSB and DCT.

References

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