

Modified SPIHT Algorithm For Biomedical Images

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Abstract- Bio-medical imaging produces a digital form of a person's body pictures in large size. So, the compression of image and data is carried out with the combination of lossy and lossless compression techniques. The lossy part helps in splitting the images into segments which are discretely encoded using a big, offline trained dictionary. The lossless part increases the entropy of the encoding between adjacent patches. Medical images are required with high quality for the diagnosis of the disease and surgery preparation. Compressing of images is required for communication purposes and to reduce memory requirements. In this project, we apply Set partitioning in Hierarchical Trees (SPIHT) for image compression on medical images. This compression is curvelet based with low computational complexity. This method achieves better compression ratios than other compression algorithms and Low PSNR.

Keywords- Curvelet Transform, Image Compression, SPHIT encoding

I. INTRODUCTION

The important information collected, handled and well explained from the human brain is the visual message. The Visual cortex consists of at least two and often a larger number in the cortical area. The main issue that arises in routine is the difficulty in exporting large amount of data. The brain images are much more important in diagnosing of diseases in the medical field. Image compression is a leading field in the Web, Conferences, Video Communication using mobiles and medical operations. It became a key factor for storing and transportation of images. Therefore image compression technique is the means of reducing the dimensional and insubstantial repetition in order to step down the number of bits. In this paper the comparison and preliminary review is done on the curvelet transform in order to achieve better compression ratio than the actual transforms. In order to compare the current transforms, a modified curvelet transform with SPIHT encoding is used..

II. LITERATURE SURVEY

A. Existing Method

There are some numbers of approaches existing and developed for image compression with different algorithm.

[1] Chandandeep Kaur have discussed the Compression of images using spiht with the use of lossless compression with better improvement. In which it shown effects on high PSNR and does not show any effect in compression ratio.

[2] Kazi Rafiqul Islam have summarized the fast ECG image compression and stored the results with low bitrate using modified SPIHT. In which it shown effect on high PSNR and low bitrate.

[3] Sure. Srikanth have discussed the efficacy of image compression for joining with Huffman encoding. The wavelet descents in SPHIT, EZW methods also looks at PSNR and bitrates of the same.

[4] Nikola Sprljan have done with the MSPIHT method for wWavelet packet images coding. He studied the next phase of wavelet packet implementation which is combination of compression scheme of SPHIT. It describes about EZW concepts based Wavelet packet Transform Coefficients.

[5] Mitul Sheth have proposed image compression of iris using distinct ways. It exhibited three methods of that rely on DWT standards. They showed the results that works through SPIHT.

III. LITERATURE REVIEW

The main objective of compression communication and repository of data. Medical imaging is important since it has a great impact on diagnosis of infection and outlining a surgery which requires a durable stockpile and valuable conveyance. Many compression schemes are available providing better compression rate but quality loss.

1D singularities can be characterized by wavelet transform. Though discrete wavelet transform has a great influence in many fields, it cannot be characterized for 2D singularities. Wavelet has some drawbacks while impending line and curve singularities. Hence Curvelet transform which resolves along smooth curves is very productive for many purposes. Curvelet transform allows to perform multiple dimensional effect on the edged objects.

Compression with wavelet transform where Haar Wavelet Transform is efficient, simplest, scalable as it can be

implemented more times with good compression ratio. It has several properties, but Set Partitioning in hierarchical trees is fully progressive which provides better quality, optically preferable and low computational effect in contrast with other algorithms such as JPEG, EBOT and BISK. Therefore SPIHT encoder is efficient approach for gray-scale images provides better quality, optically preferable and less computing fallout in contrast with other format algorithms like EBOT, JPEG, and BISK. Therefore SPIHT encoder is efficient approach for gray-scale defined.

A. Curvelet Transform

Curvelet Transform is a function used to represent the curve as the function of varying length and width which obeys scaling law. Curvelet Transform functions based on the process of splitting the image into sub-bands at first phase, which means, splitting the image of the object into two unique parts and each part is then subjected to the Local Ridgelet Transform.

In the modified Curvelet Transform the process of decomposing is done, where the input Image is split into 2 parts. The sub-bands are then analyzed by ridgelet transform. Decomposing with Curvelet transform results in high compression ratio. The given input Image is decomposed with the help of a modified Curvelet Transform depend on the upcoming four steps:

- Sub-band Decomposition
- Smooth Partitioning
- Renormalization
- Ridgelet Analysis

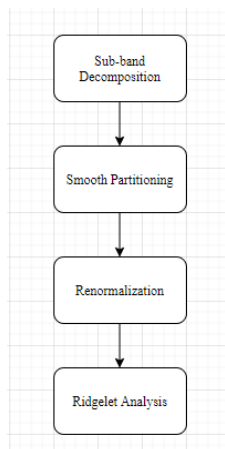


Fig. 1. Block diagram of Curvelet Transform

B. SPHIT Encoding

The process is coding the coefficients and achieve the results by decomposition of modified curvelet. Though there are numerous algorithm for encoding the most effective algorithm for image compression that provides effect on factors such as better virtue of image, clear vision, fastest way to encode and decode, Low-bit rate performance and Low computational effect. SPHIT algorithm is,

$$S_n(\Gamma) = \begin{cases} 1, & \max(i,j) \in \Gamma \{ |c_{ij}| \} \geq 2^n \\ 0, & \text{otherwise} \end{cases}$$

Based on the threshold value they are classified into three basic sets as, List of Insignificant sets (LIS), List of Insignificant Pixels (LIP), List of Significant Pixels (LSP), respectively.

C. Algorithm

O(I,j), Set of Coordinates of all Offspring nodes (I,j): children only.

D(I,j), Set of Coordinates of all Descendent nodes: children, grand children.

H(I,j), Set of all Tree Roots which is the node with the Highest Pyramid level: Parents.

L(I,j), D(I,j)-O(I,j): all Descendents except the Offspring: Grand children.

IV. PEAK SIGNAL TO NOISE RATIO

PSNR helps in estimating the ratio between the Maximum power of a Signal and the corrupted noise signal that affects the originality of the Signal representation. It is also a measurement to measure the clarity and quality of the image. The Higher the PSNR value represents the reconstructed quality of Image is high, and the Lower the PSNR value denotes the reconstructed quality of Image is low. MSE helps in verifying the Mean Square Error of the Image.

$$MSE = \frac{\text{sum}(\text{sum}(\text{squaredErrorImage}))}{(\text{row} \times \text{columns})}$$

$$PSNR = 10 \log_{10}(255^2 / MSE)$$

A. Compression Ratio

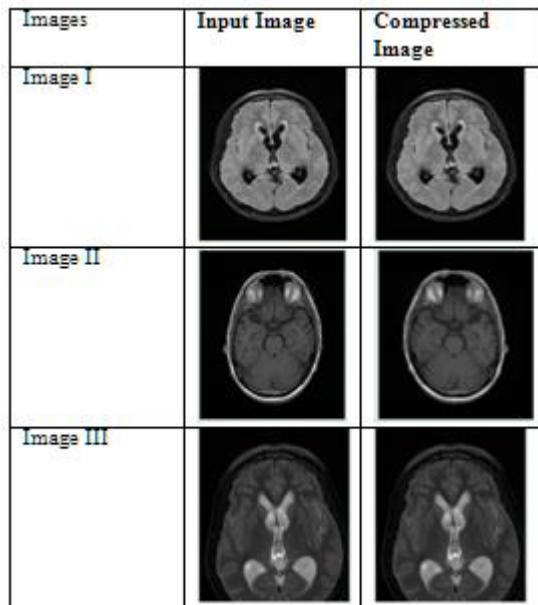
Compression Ratio (CR) Is described as the ratio between the size of the Original Image and size of the Compressed Image.

$$CR = \frac{\text{original Image}}{\text{Compressed Image}}$$

V. PERFORMANCE AND SIMULATION RESULT

The performance is estimated for distinct transforms using the medical images of the brain. From the existing model the quality and efficiency of the image has been obtained using various metrics.

The images below represent the Human brain before compression and after compression



A. Figures and Tables

TABLE I.

	Compression Ratio		PSNR	
	Proposed Scheme	Existing Scheme	Proposed Scheme	Existing Scheme
Image 1	1.2698	1.2378	26.25	31.08
Image 2	1.5675	1.4591	24.91	33.09
Image 3	1.6452	1.2891	24.59	32.68
Image 4	1.164	1.123	24.29	35.22

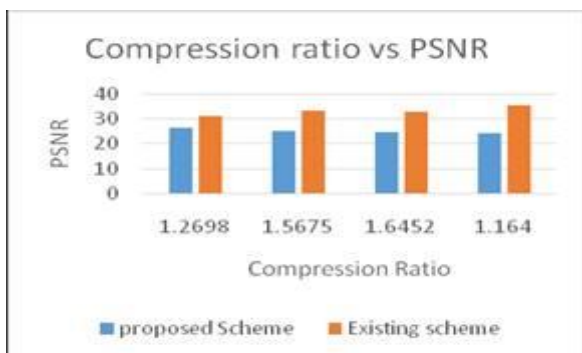


Fig.2.Plot of Compression Ratio and PSNR

Table I represents the comparison between Compression Ratio and PSNR of the existing and the proposed scheme which is also graphically pictured below. From the graph it is understood that the CR and PSNR values of the proposed scheme are optimized than the existing scheme.

TABLE II.

	Compression Ratio		MSE	
	Proposed Scheme	Existing Scheme	Proposed Scheme	Existing Scheme
Image 1	1.2698	1.2378	154.017	50.670
Image 2	1.5675	1.4591	209.553	31.909
Image 3	1.6452	1.2891	225.493	35.075
Image 4	1.164	1.123	241.665	19.507

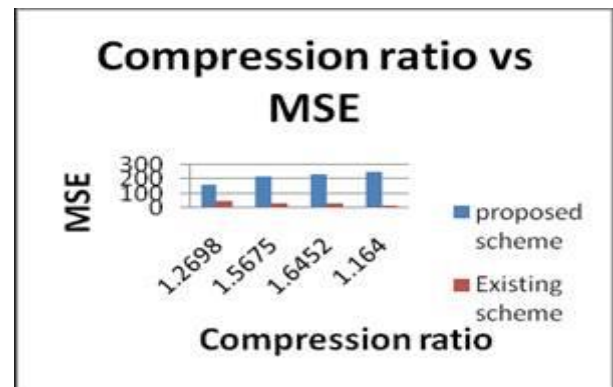


Fig.2.Plot of Compression ratio and MSE

Table II shows the comparison results of compression ratio and MSE and the above graph depicts the table. From the graph it is clear that existing scheme less efficient than the latter.

VI. BLOCK DIAGRAM

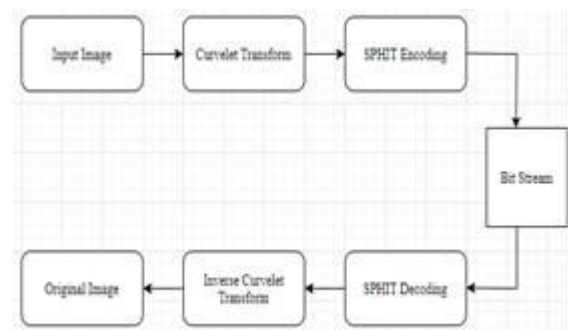


Fig. 3. Block Diagram of modified SPHIT algorithm

VII. CONCLUSION

Thus the proposed scheme was compared with the existing scheme and experimentally interpreted. The

interpretation made clear on Compression ratio and PSNR of the image. Experimental evaluation shows that SPHIT carries off an immeasurable increase in high Peak signal to noise ratio(PSNR) value and compression ratio(CR) than the other techniques. As the obtained results match all the needs of curvelet transform, precisely it is clarified, adaptive image coding are operated in the space where there are no quality loss and deterioration of the images.

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