

# Analysis of Wavelet Transform for Image Compression Using SPITH Algorithm

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**Abstract** -Last few years there has been astronomical increase in the usage of computers. In digital cameras, one of the most common uses has been the storage, manipulation, and transfer of digital images. Image compression is one of the important applications in data compression on its image. Image data requires huge amount of disk space and large bandwidths for transmission. In multimedia application, most of the images are in color and color images contain more data redundancy and require a huge amount of storage space. This project presents a new approach for compression of color images and gray scale using SPIHT algorithm along with wavelet transform. In this project we produce a new technique as adaptive lifting DWT. An Adaptive Lifting DWT that locally adapts the filtering directions to image content based on directional lifting. This technique using the new algorithm that detects all the blocks in a given image to decide whether the block is homogenous or heterogeneous block. For homogeneous block, the simple Discrete Wavelet Transform (DWT) is used and for the heterogeneous block, Lifting Wavelet Transform is used. In this technique image quality is measured objectively, using peak signal noise ratio or picture quality scale using perceived image quality.

**Key Words:** SPIHT, Wavelet, Lifting scheme, Entropy Coding, Image Compression

## 1. INTRODUCTION

Digital image compression is now essential. Internet teleconferencing, High Definition Television (HDTV), satellite communications and digital storage of images will not be feasible without a high degree of compression. Wavelets became popular in past few years in mathematics and digital signal processing area because of their ability to effectively represent and analyze data. Typical application of wavelets in digital signal processing is image compression. The compression provides to reduce the cost of storage and increase the speed of transmission. Image compression is used to minimize the size in bytes of a graphics file without degrading the quality of the image. There are two types of image compression is present lossy and loss less[18]. The lossy type aims to reduce the bits required for storing or transmitting an image without considering

the image resolution much and the lossless type of image compression focuses on preserving the quality of the compressed image so that it is same as the original image.

Image compression algorithms based on Discrete Wavelet Transform (DWT), such as Embedded Zero Wavelet (EZW)[3] which produces excellent compression performance, both in terms of statistical peak signal to noise ratio (PSNR) and subjective human perception of the reconstructed image. Said and Pearlman further enhanced the performance of EZW by presenting a more efficient and faster implementation called set partitioning in hierarchical trees. SPIHT is one of the best algorithms in terms of the peak signal-to-noise ratio (PSNR) and execution time. In this paper, we propose an algorithm for digital image compression based on lifting base wavelet transform coupled with SPIHT (Set Partition in Hierarchical Trees) coding algorithm, of which we applied the lifting structure to improve the drawbacks of conventional wavelet transform. We compared the results with various wavelet based compression algorithm. Experimental results show that the proposed algorithm is superior to traditional methods for all tested images at low bit rate. Our algorithm provides better PSNR, Quality factor and MSSIM values for medical images only at low bit rate. Compressed image can be represent in various format such as GIF, JPG, BMP and PNG.

## 2. Methodology

**A. SPIHT:** SPIHT is Set Partitioning In Hierarchical Trees. It is the wavelet based compression coder. It divides the wavelet into Spatial Orientation Trees. SPIHT codes a wavelet by transmitting information about the significance of a pixel. It is a method of coding and decoding the wavelet transform of an image.

1. Implementation of SPIHT: The basic principle is progressive coding which process the image respectively to a lowering threshold. First step, the original image is decomposed into sub bands. Then the method finds the maximum iteration number. Second, the method puts the DWT coefficients into a sorting pass that finds the significance coefficients in all coefficients and encodes the sign of these significance coefficients. Third, the significance coefficients that can be found in the sorting pass are put into the refinement pass that uses two bits to exactly reconstruct value for approaching to real value [1]. The result is in the form of a bit stream. It has three lists to store the values. They are List of Insignificant Pixels (LIP), List of Significant Pixels (LSP), and List of Insignificant Sets (LIS).

**B. MSPIHT:** In SPIHT, the usage of three temporary lists are quite memory consuming. In addition, during coding the elements in the lists are often inserted or deleted, thus, greatly increase the coding time with the expansion of the lists. Thus Modified SPIHT algorithm varies from SPIHT algorithm by the way in which the subsets are partitioned and significant information is conveyed. In the proposed MSPIHT algorithm, the sorting pass and the refinement pass are combined as one scan pass. The lists LIP and LSP are realized in one RAM module and consequently the area information is stored. According to the characteristic of DWT, if a coefficient is significant at a certain threshold then its neighbors will be significant at the next threshold with a high probability. So we can scan the neighbors of significant coefficients in advance, so that more significant coefficients can be encoded at a specified bit rates. The wavelet Lifting Scheme is a method for decomposing wavelet transforms into a set of stages. Lifting scheme algorithms have the advantage that they do not require temporary arrays in the calculations steps and have less computation.

### A. Splitting

In this stage the input signal is divided into two disjoint sets, the odd ( $X[2n+1]$ ) and the even samples ( $X[2n]$ ).

### B. Lifting

In this module, the prediction operation P is used to estimate  $X_0(n)$  from  $X_e(n)$  and results in an error signal  $d(n)$ . Then we update  $d(n)$  by applying it to the update operation U, and the resulting signal is combined with  $X_e(n)$  to  $S(n)$  estimate, which represents the smooth part of the original signal.

### C. Scaling

A normalization factor is applied to  $d(n)$  and  $s(n)$ , respectively. In the even-indexed part  $S(n)$  is multiplied by a normalization factor  $K_e$  to produce the wavelet sub-band  $XL1$ . Similarly in the odd-index part the error signal  $d(n)$  is multiplied by  $K_o$  to obtain the wavelet sub band  $XH1$ . The output result is  $XL1$  and  $XH1$  by using the lifting-based WT are the same as those of using the convolution approach for the same input. For lifting implementation, the CDF 9/7 wavelet filter pair can be factorized into a sequence of primal and dual lifting. The most efficient factorization of the poly phase matrix for the 9/7.

## 3. Performance Of Image Compression Scheme

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K: the 4x4 block to be processed.
L: the 2x2 sub-block of K.
n: the bit-plane number
1: Refinement Pass (RP):
2: for each L in K such that  $S_{n+1}(L) = 1$  do
3:   for each w(i) in L do
4:     output  $S_n(w(i))$  (= the nth bit of  $|w(i)|$ ); # magnitude bit
5:     if  $S_{n+1}(w(i)) = 0 \wedge S_n(w(i)) = 1$  then
6:       output the sign of  $w(i)$ ; # sign bit
7:   set  $S_n(L)$  to 1
8:
9: Sorting Pass (SP):
10: if  $S_{n+1}(K \cup D(K)) = 0 \wedge \sim \{parent(K) \wedge S_n(parent(K)) = 0\}$ 
11:   then
12:     output  $S_n(K \cup D(K))$ ; # sorting bit
13:   if  $S_n(K \cup D(K)) = 1$  then
14:     for each L in K such that  $S_{n+1}(L) = 0$  do
15:       output  $S_n(L \cup D(L))$ ; # sorting bit
16:       if  $S_n(L \cup D(L)) = 0$  then
17:         set  $S_n(L)$  to 0
18:       else
19:         set  $S_n(L)$  to 1
20:   else
21:     for each L in K such that  $S_{n+1}(L) = 0$  do
22:       set  $S_n(L)$  to 0
23:
24: First Refinement Pass (FRP):
25: for each L in K such that  $S_n(L) = 1$  and  $S_{n+1}(L) = 0$  do
26:   set  $S_n(L)$  to 1
27:   for each w(i) in L do
28:     output the nth bit of  $|w(i)|$  # magnitude bit
29:     if  $S_n(w(i)) = 1$  then
30:       output the sign of  $w(i)$  # sign bit

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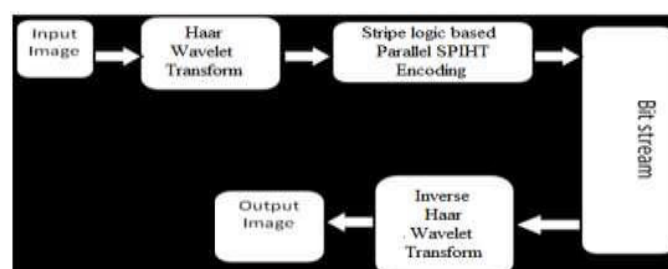


Fig 1. Block diagram of the entire process

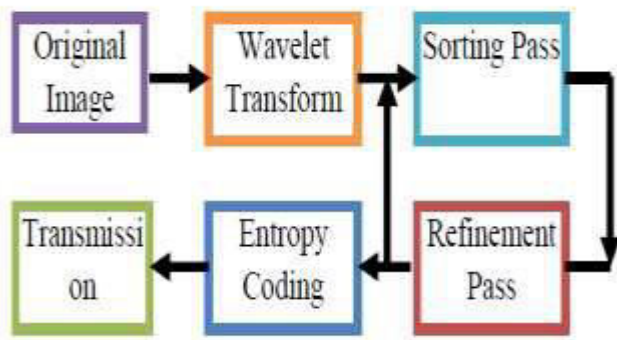


Fig.2Flowchart of SPIHT Algorithm

7. Mr.Vipin V and Miranda Mathews, “FPGA Implementation of Image Compression using SPIHT technique” , IJAREEIE January 2014, PP. 7001-7007.

8.Ping Liu, Guanfeng Li, “An Improved SPIHT Algorithm for Image Compression in Low Bit Rate”, Scientific Research, September 2013, PP. 245-248.

### 3. CONCLUSIONS

The proposed method is based on HWT and SPS technique for medical images is presented for real time applications. This technique reduces the memory requirements, enhances the speed and retains all the advantages of embedded coding. The utilization of this method can reduce the MSE or enhance the PSNR values which are the important performance parameters.HWT makes itself a standard technique for its high efficiency. SPIHT which is a powerful technique for still images offers various good characteristics like, good image quality, high PSNR, fast coding and decoding etc., along with HWT technique provides an efficient design. Experimental results demonstrate that the PSNR value obtained by this method may reach up to 83.6dB . As a future enhancement, the work can be extended for designing higher resolution images of size 512×512, 1024×1024 etc. Number of decomposition levels can be increased that may provide still better compression. Other coding technique can be used instead of bit-plane coding since it is not relevant for decoding to obtain still better results.

### REFERENCES

1. YongseokJin, Hyuk- Jea Lee, “A Block -based Pass-Parallel SPIHT algorithm”, IEEE, July 2012, PP. 1064-1075.BPS
2. Kai Liu, JieGuo and Evgeniy Belyaev, “VLSI Architecture of Arithmetic coder used in SPIHT”, IEEE, Vol 20, April 2012, PP. 697-710.neighbor
3. K. RaghunadBhattar, K.R, Ramakrishnan and Dasguptha K.S, “Stripe based embedded coding of wavelet coefficients for large images”, IISC, 2013. [I92]
4. B. PritiBramhankar and B. Sachin, “Lifting based wavelet transform along with SPIHT algorithm used for image compression”, IJMTER , July 2013, PP. 2033- 2037. [liftin]
5. Yin-hua Wu, Long-xuJin and Hong-jiang Tao, “An improved fast parallel SPIHT algorithm and its FPGA implementation”, IEEE, 2010, PP. 191-195. [2]
6. Youngseop Kim and William A. Pearlman , “Stripe-Based SpihtLossy Compression of Volumetric Medical Images for Low Memory Usage and Uniform Reconstruction Quality”, IEEE, 2000, PP. 652-655