

## International Journal of Control Theory and Applications

ISSN: 0974-5572

© International Science Press

Volume 10 • Number 15 • 2017

# An Improved Daubechies Wavelet based Image Compression methods

#### Puja Dipak Saraf<sup>1</sup>

<sup>1</sup> Department of Computer Engineering RCPIT, Shirpur, Maharashtra, India, Email: pujasaraf20@gmail.com

*Abstract:* Real time transmission of images through handheld mobile or convenient devices requires an image coding algorithm to reduce the redundancies among the image in the form of pixels at the very low bit rate. Large number of successful wavelet based image coding algorithm has been proposed in the literature. These algorithms are mainly classified into zero tree and zero block approaches. Zero tree algorithm make use of inter sub bands correlations existing among different sub bands of an image. The inefficiency of zero block lies in that they do not take care of inter sub bands correlation existing among different sub bands. Here present an implementation of a wavelet coder based algorithm. The above methods are used in many applications for compression and decompression of an image. The techniques are compared by using daubechies wavelet family for performance measuring parameter like higher compression ratio, bits per pixel and the amount of time required at every stage after the decomposition of an image. The comparative analysis was done using the four methods with the results are mentioned in this paper.

Keywords: Wavelet, DWT, CR, ROI, decompositions

#### 1. INTRODUCTION

Uncompressed multimedia data needs significant storage capacity and transmission bandwidth. The data are in the form of graphics, audio, video and image. These types of data contains large amount of information in the form of pixel. During transmission it is necessary to reduce this information by compressing the data. Large amount of data can't be stored if there is low storage capacity. The compression offers a means to reduce the cost of storage and increase the speed of transmission. One important development happening in the whole digital industry is the increase in the image size and resolution. The image compression is used to minimize the size of in bytes of a graphics file without corrupting the quality of the image [1]. The key concept of compression is irrelevance and redundancy reduction. Removing duplication from the original image is carried out by redundancy reduction whereas the irrelevancy reductions omits the part of the signal which cannot be noticed by the signal receives like Human Visual System (HVS) [2] The three kinds of redundancy are as follows a) Spatial Redundancy between neighboring pixel values, b) Spectral Redundancy between different color planes or spectral bands, c) Temporal Redundancy between adjacent frames in a sequence of images (in video applications). The image compression research aims at reducing the number of bits needed to represent an image by removing the spatial

#### Puja Dipak Saraf

and spectral redundancies as much as possible [3]. Basic types of image compression are lossless and lossy. Lossless compression works by compressing the overall image without removing any of the image's details. Lossy compression works by removing image's details but not in such a way that it is apparent to the viewer [4]. The rest of the paper is organized as follows. Literature Survey is explained in section II. Methodology is explained in section III. Experimental results are given in section IV. Concluding remarks are given in section V.

#### 2. LITERATURE SURVEY

Over the past decade's variety of novel and sophisticated wavelet based and cosine based image coding schemes have been developed. For lossless methods these includes Run length encoding schemes proposed by Robert Zigon were employed in the transmission of television signals. It is well suited to palette based bitmapped images such as computer icons. It does not work well at all on continuous tone images such as photographs while JPEG uses it fairly well on the well-organized that remains after transform and quantizing image block. RLE is a lossless compression method, but it is partial to 2:1 or 3:1. RLE is most effective with less points of gray [5]. David Huffman proposed the Huffman code particular type of optimal prefix code that is commonly used for lossless data compression. The output from Huffman algorithm can be viewed as variable code table for encoding a source symbol (such a character in file). The algorithm arises the table from the predictable chance or occurrence (weight) for each possible value of source symbol. But this code has some limitations that are Huffman code is optimal only if the exact probability circulation of the source symbols is known. Each symbol is encoded in an integer number of bits. Huffman coding is not efficient to adapt with the changing source statistics. The length of the codes of the smallest amount probable image might be terribly massive to store into one word or basic storage unit during a system [8]. Lempel-Ziv and Welch proposed dictionary based algorithms scan a file for sequences of information that occur quite once. These sequences are then stored in a dictionary and within the compressed file, locations are set where ever monotonous data occurred. LZW compression replaces strings of characters with single codes. The code that the LZW algorithmic program outputs may be of any random length, however it primarily have a lot of bits in it than one character. The first 256 codes (when using eight bit characters) are firstly allocated to the regular character set. The remaining codes are assigned to the strings as the algorithm proceeds [9]. The basic Joint Photographic Expert's Group standard for coding both monochrome and color still images was introduced in 1991 as a result of collaborative activity between working and study groups of the ISO and CCITT (now ITU-T) and was intended to provide reconstruction of conventional ITU-R recommendation 601 format 4:2:2 pictures at realistic quality in the 0.25-0.5 bit/element region through to no perceptible degradation at about 1.5 bit/element proposed by Clarke in 1995 and then by Rao and Hwang, 1996[6]. Moving picture experts Group (MPEG) MPEG was originally created as a standard for very low bit rate coding of limited complexity audiovisual material. The scope was later extended to supporting new functionalities such as content-based interactivity, universal access, and high-compression coding of general material for a wide bit-rate range. It also emphasizes flexibility and extensibility. Comparing MPEG video coding with earlier standards, the major difference lies in the representation and compression of the shape information. In addition, one activity that distinguishes MPEG from the conventional video coding standards is the synthetic and natural hybrid coding. Comparatively the MPEG structure is more complex. The target technologies studied include face animation, coding and representation of a 2D dynamic mesh, wavelet-based static texture coding, viewdependent scalability, and 3D geometry compression [7]. J.M.Shapiro Proposed first algorithms to show the full power of wavelet-based image compression known as EZW. The EZW encoder is based on progressive encoding to compress an image into a bit stream with growing correctness. This means that when extra bits are added to the stream, the decoded image will contain more detail [10]. Amir Said and Perlman Proposed the SPIHT methods consecutively improved the EZW algorithm by covering this coding scheme. This new coding scheme provided an even better performance than the improved version of EZW. It is a powerful image compression algorithm that crops an embedded bit stream after which the greatest recreated images in the mean square error sense can be extracted at various bit rates. Some of the best results-highest PSNR values for given compression

ratios for a wide variety of images [11]. Tian and wells proposed Wavelet Difference Reduction methods finds the defects of SPIHT is that it only indirectly detects the location of substantial coefficients. This makes it tough to achieve operations which depend on the position of significant transform values, such as region choice on compressed data. Region choice, also known as region of interest (ROI), means a portion of a compressed image that requires increased resolution [12]-[13]. James Walker Proposed Adaptively Scanned Wavelet Difference Reduction methods the Modifies the scanning order used by WDR in order to achieve better performance and predict Locations of new significant values. It also allows the ROI capability [14]. David Taubman proposed *EBCOT algorithm* 1998 The EBCOT algorithm uses a wavelet transform to generate the sub band coefficients which are then quantized and coded. While the regular dyadic wavelet decomposition is standard, other "packet" decompositions are also supported and occasionally preferable. The original image is signified in terms of a gathering of sub bands, which may be organized into growing resolution levels [15].

## **3. METHODOLOGY**

Wavelet based image compression methods has achieve more acceptance because of their overlapping flora which minimizes the blocking artifacts and multi-resolution appeal, outcomes the high superiority reconstructed images. Wavelet based techniques provides significant improvements in picture quality at higher compression ratios. Also, at higher compression ratios, wavelet coding methods reduce abundant extra gracefully than the DCT method. Wavelet based compression techniques allow the addition of various compression methods into one and hence, a compression ratio of up to 200:1 is obtainable. Different wavelet based methods for image compression have been developed and implemented over the past few years. These include Adaptively Scanned Wavelet Difference Reduction (ASWDR), Embedded Zero Tree Wavelet (EZW), Set-Partitioning in Hierarchical Trees (SPIHT), Wavelet Difference Reduction (WDR) etc. Here the algorithm uses Compression techniques by performing the following the steps:

- 1) Image transformation is applied to transform the image from one domain to another domain.
- 2) Wavelet Transform is applied to the image then the image is divided into low and high frequency domain, again the part of image which contains the more information again the wavelet transform is applied to that level.
- 3) Performs the levels of decompositions for getting the more and more clear image by reducing the redundant information from image.
- 4) Wavelet reconstructions is applied for getting the original image.

## 4. EXPERIMENT AND RESULT

Here we have performed the experiment on Lena Image for measuring the different performance metrics like Compression Ratio, bits per pixel and the time for executing the algorithm. The Compression ratio is calculated by size of original image and image after the compression. Bits per pixel is nothing but the number of bits of information stored per pixel of an image. For that purpose different Daubechies wavelet are used that includes db2, db4, db6 and db10 wavelet. With every wavelet applying to an image by increasing the levels of decomposition. At every stage we are getting the output image.

Here the lena image is obtained after applying the Discrete Wavelet Transform i.e if in any part of the image where the maximum energy is stored due that part the again we apply the DWT. This process is continued untill upto we get the more compression ratio with quality and bits per pixel. By removing the redundant information from that image that will be easily transferable and less amount of space will be required. The results are explained at every stage with db2, db4, db6 and db10 wavelets. We have to measure that at which stage we are getting the good results and with bpp and the time to execute the algorithm. As described in the in

Puja Dipak Saraf



Table 1         Performance Metrics with Daubechies Wavelet family         Performance Metrics with Daubechies wavelet family											
CR	BPP	Time	CR	BPP	Time	CR	BPP	Time	CR	BPP	Time
10.54	0.84	64.5	9.6	0.77	28.71	9.65	0.77	18.17	12.55	1	18.99
9.74	0.78	29.04	8.87	0.071	63.45	8.84	0.71	60.8	11.32	0.91	32.55
3.25	0.26	18.21	2.93	0.23	24.17	2.91	0.23	26.66	3.74	0.3	25.55
10.9	0.87	21.67	9.84	0.79	21.51	9.86	0.79	18.11	13.33	1.07	17.49
	CR 10.54 9.74 3.25 10.9	db2           CR         BPP           10.54         0.84           9.74         0.78           3.25         0.26           10.9         0.87	Perf           Perf           db2         Perf           CR         BPP         Time           10.54         0.84         64.5           9.74         0.78         29.04           3.25         0.26         18.21           10.9         0.87         21.67	Performance           Performance           db2         Performance           CR         BPP         Time         CR           10.54         0.84         64.5         9.6           9.74         0.78         29.04         8.87           3.25         0.26         18.21         2.93           10.9         0.87         21.67         9.84	Performance Metrics of Performance Metrics of Performance Metrics of Adda and Addition and Additional and Additionand Additi Additionand Additiona and Additional and Additional and	Table 1           Performance Metrics with Daube           Performance Metrics with Daube           db2         db4           CR         BPP         Time         CR         BPP         Time           10.54         0.84         64.5         9.6         0.77         28.71           9.74         0.78         29.04         8.87         0.071         63.45           3.25         0.26         18.21         2.93         0.23         24.17           10.9         0.87         21.67         9.84         0.79         21.51	Table 1           Performance Metrics with Daubechies W           Performance Metrics with Daubechies wa           db2         db4         CR         BPP         Time         CR           10.54         0.84         64.5         9.6         0.77         28.71         9.65           9.74         0.78         29.04         8.87         0.071         63.45         8.84           3.25         0.26         18.21         2.93         0.23         24.17         2.91           10.9         0.87         21.67         9.84         0.79         21.51         9.86	Table 1         Performance Metrics with Daubechies Wavelet fami         Performance Metrics with Daubechies wavelet fami         db2       db4       db6         CR       BPP       Time       CR       BPP       Time       CR       BPP         10.54       0.84       64.5       9.6       0.77       28.71       9.65       0.77         9.74       0.78       29.04       8.87       0.071       63.45       8.84       0.71         3.25       0.26       18.21       2.93       0.23       24.17       2.91       0.23         10.9       0.87       21.67       9.84       0.79       21.51       9.86       0.79	Table 1         Performance Metrics with Daubechies Wavelet family         Performance Metrics with Daubechies wavelet family         db2       db4       db6         CR       BPP       Time       CR       BPP       Time       CR       BPP       Time         10.54       0.84       64.5       9.6       0.77       28.71       9.65       0.77       18.17         9.74       0.78       29.04       8.87       0.071       63.45       8.84       0.71       60.8         3.25       0.26       18.21       2.93       0.23       24.17       2.91       0.23       26.66         10.9       0.87       21.67       9.84       0.79       21.51       9.86       0.79       18.11	Table 1         Performance Metrics with Daubechies Wavelet family         Performance Metrics with Daubechies wavelet family         db2       db4       db6         CR       BPP       Time       CR       BPP       Time       CR         10.54       0.84       64.5       9.6       0.77       28.71       9.65       0.77       18.17       12.55         9.74       0.78       29.04       8.87       0.071       63.45       8.84       0.71       60.8       11.32         3.25       0.26       18.21       2.93       0.23       24.17       2.91       0.23       26.66       3.74         10.9       0.87       21.67       9.84       0.79       21.51       9.86       0.79       18.11       13.33	Table 1         Performance Metrics with Daubechies Wavelet family         Performance Metrics with Daubechies wavelet family         Daubechies wavelet family         db2       db4       db6       db10         CR       BPP       Time       CR       BPP       Time       CR       BPP         10.54       0.84       64.5       9.6       0.77       28.71       9.65       0.77       18.17       12.55       1         9.74       0.78       29.04       8.87       0.071       63.45       8.84       0.71       60.8       11.32       0.91         3.25       0.26       18.21       2.93       0.23       24.17       2.91       0.23       26.66       3.74       0.3         10.9       0.87       21.67       9.84       0.79       21.51       9.86       0.79       18.11       13.33       1.07

 WDR
 10.9
 0.87 21.67 9.84 0.79 21.51 9.86 0.79 18.11 13.33 1.07 17.49 

 the Table 1 if we used db2 ,db4,db6 and db10 wavelet with WDR algorithm we are getting the good CR ratio.

the Table 1 if we used db2,db4,db6 and db10 wavelet with WDR algorithm we are getting the good CR ratio. But sometimes there may be chances of degrading the image quality by getting the more and more compression ratio.

## 5. CONCLUSION

100

200

300

400

500

100

Here different image compression techniques and its implementation is given. It is seen that these techniques of image compression offers various advantages as it is fully embedded codec and very simple powerful error correction and progressive image transmission techniques. As these techniques are wavelet based image compression so the discrete wavelet transform plays a vital role. Here an adaptively scanned wavelet difference reduction methods yields an embedded codec, comparable in performance to the widely used SPIHT codec, but which has ROI capability and it is better at preserving edge details. We have implemented and tested the quality

assessment of image compression with various algorithms of wavelet families. With db10 family EZW, WDR and ASWDR perform well in terms of compression ratio whereas the SPIHT gives the better quality of the image but with less compression ratio than the WDR, ASWDR and EZW.

#### ACKNOWLEDGMENT

The authors would like to thanks for Vice Chancellor Research Motivational Schemes, North Maharashtra University, Jalgaon for giving valuable Opportunity to us. Also thankful to R.C.Patel Institute of Technology for giving the moral support.

#### REFERENCES

- S.P.Raja, Dr.A.Suruliandi, "Analysis of Efficient Wavelet based Image Compression Techniques", 2010 Second Iternational conference on Computing, Communication and Networking Technologies.
- [2] Parvinder Kaur, "Compression Using Fractional Fourier Transform" A Thesis submitted in the partial fulfillment of the requirements for the award of the degree of Master of Engineering In Electronics and Communication.
- [3] VimalRathinasamy, IyyapanDhasarathan, Tang Chi, "Wavelet Based SPIHT Compression for DICOM Images", Linnaeus University, School of Computer Science, Physics and Mathematics, 2011.
- [4] SubhasisSaha, "Image compression-from DCT to Wavelets: a review", New York, NY, USA, Volume 6 Issue 3, Volume 6 Issue 3, March 2000.
- [5] Martin H, Weik, Computer Science and Communication Dictionary, Volume1, p. 129 2000,
- [6] Roger J. Clarke, "Image and Video Compression: A Survey", Department of Computing and Electrical Engineering, Heriot-Watt University, Riccarton, Edinburgh EH14 4 AS, Scotland.
- [7] Ling Guan, Sun-Yuan Kung, Jan Larsen, "MULTIMEDIA IMAGE and VIDEO PROCESSING", CRC Press, ISBN 0-8493-3492-6, Boca Raton London New York Washington, D.C, 2000.
- [8] Gonzalez, R.C And Woods, R.E, "Fundamentals Of Digital Image Processing", 2nd ed, prentice Hall upper saddle river, NJ, 2002.
- [9] Sawsan A. Abu Taleb, Hossam M.J. Musafa, Asma'a M. Khtoom, Islah K. Gharaybih, "Improving LZW Image Compression", European Journal of Scientific Research, ISSN 1450-216X Vol.44 No.3, pp.502-509, 2010.
- [10] Jerome M Shapiro Embedded Image coding Using zerotrees of wavelets coefficients", IEEE Transactions on signal processing, Vol 41, no. 12, pp. 3445-3462, Dec 1993.
- [11] Amir Said, and Pearlman, W. A. A New, Fast, and Efficient Image Codec Based on Set Partitioning in Hierarchical Trees, IEEE Trans. Circuit and systems for Video Technology, vol. 6, no. 3, pp. 243-250, June 1996.
- [12] Wallace, G.K., The JPEG still picture compression Standard, Comm. of the ACM, 34(4), 30–44, 1991.
- [13] Puja D Saraf, DeeptiSisodia, Amit Sinhal and Shiv, "Comparisons of wavelets based image compression methods", WJST 2 (3), 10-13, 2012.
- [14] Walker, J.S., Alossy image codec based on adaptively scanned wavelet difference reduction, Optical Engineering, in press.
- [15] ]Taubman, D.High Performance Scalable Image Compression with EBCOT, submitted to IEEE Transactions on. Image Processing, Mar.1999, http://maestro.ee.unsw.edu.au/~taubman/activities/pre prints/ebcot.zip