# Robust Blind Complex Double Haar Wavelet Transform Based Watermarking Algorithm for Digital Images

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Abstract—Dual-Tree Complex Wavelet Transform is relatively a recent improvement of the Discrete Wavelet Transform (DWT) with important additional properties like shift invariant and directionality. In this paper, we propose a blind watermarking scheme based on Complex Double Haar Wavelet Transform (CDHWT). Single level CDHWT is applied on host image, it decompose the original image into dual nine subbands and single level DHWT is applied to the binary watermark, it decompose the binary image into nine subbands. Eigen values of the selected subband are replaced by the Eigen values of the selected subband of binary watermark which is multiplied by an appropriate strength factor. There is no need of host image and original watermark to extract the watermark from the test image. An experimental result showed that the proposed scheme achieved very high imperceptibility and Robustness against various image processing attacks like JPEG compression, low pass filtering, median filtering, addition of noise, rotation, cropping and histogram equalization etc.

Index Terms—M-Channel filter bank, DHWT, CDHWT, SVD and digital watermarking.

# I. INTRODUCTION

Now a days, digital watermarking plays a major role in multimedia security tools. Because any form of media like image, audio, video and data can be watermarked. Various watermarking methods are proposed for different applications. Novel watermarking techniques are classified into two types, Spatial domain technique and Transform domain technique. In Spatial domain technique [1]-[3], pixel value is directly modified to embed the secret information. In Transform domain technique, Original image is transformed into transform coefficients by using various popular transforms like DCT [4], DFT [5] and DWT [6]-[10] etc. Transform coefficients are modified to embed the secret information. Transform domain technique achieves more robustness as compared to spatial domain technique but it needs more computational complexity.

Due to its multi resolution property, wavelet transform has more application in watermarking on images [6], [7]. Various types of wavelet transforms are employed for different kinds of image watermarking [6]-[10]. Embedding of secret information in different frequency domains has its own advantages and disadvantages. The low-frequency embedding of the watermark increases the robustness with respect to image distortions that have low pass characteristics

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like filtering, lossy compression, geometrical distortions. On the other hand, low-frequency watermarks are more sensitive to modifications of the histogram, such as contrast/brightness adjustment, gamma correction, histogram equalization, and cropping. Watermarks inserted in high frequencies are typically less robust to low-pass filtering, lossy compression and small geometric deformations of the image. But, they are extremely robust with respect to noise adding, nonlinear deformations of the gray scale. To compromise between these two, mid frequencies are selected to embed the watermark.

Wavelet transform is a very popular technique in image transform. Various watermarking methods are proposed in wavelet domain due to their excellence of multi resolution property. Byun et al. [16] proposed a watermarking method using quantization and statistical characteristics of wavelet transform. Wang et al. [34] proposed a wavelet tree based blind watermarking scheme. Jiang et al. [19] proposed a blind watermarking scheme based on 4-band wavelet transform. An Integer Wavelet Based Logo-watermarking Scheme was proposed by yuan et al. [31]. Preprocessed watermark is embedded in the low and high frequency subbands. Mahmood et al [32] proposed a semi blind watermarking scheme using image denoiseing based on DWT. Li [35] et al proposed wavelet tree quantization based watermarking scheme robust to geometric attacks like rotation, scaling and cropping. Peng et al [15] proposes a blind image watermarking scheme using wavelet trees quantization. Wei et al [18] proposed a blind watermarking algorithm based on the significant difference of wavelet coefficient quantization.

However, Scalar wavelets are generated by one scaling function [19]. It does not support orthogonality and symmetry simultaneously. Multiwavelets which have more than one scaling function can simultaneously provide better reconstruction while preserving length. Good performance at the boundaries and a high order of approximation are added features. Thus, multiwavelet provides superior performance for image processing applications, compared with scalar wavelets [19]. The Haar wavelet transform consistently outperform the more complex ones when using non-colored watermark [10]. At the same time, DWT suffers from oscillations, shift invariance, aliasing and lack of directionality as it is based on real valued oscillating wavelets.

However, Fourier transform does not suffer from this problem because it is based on complex valued oscillating sinusoids [19]. Therefore, shortcomings of real valued DWT were overcome by the Complex Wavelet Transform [20]. In recent years, Complex Wavelet Transform took more attention in image transformation as well as image

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watermarking [21],[22]. In this paper we propose a Complex Double Haar Wavelet Transform based watermarking scheme. Thus, it combines the advantages of both multiwavelet and complex representation of an image. Section II discusses the M-channel Filter bank and DHWT (Double Haar Wavelet Transform). Section III discusses Complex wavelet transform and CDHWT. Section IV discusses the proposed embedding and extraction algorithm using CDHWT. Section V discusses the experimental result of the proposed algorithm for different gray scale images and comparison with existing DWT based methods followed by conclusion in section VI.

#### II. M-CHANNEL FILTER BANK AND DHWT

Multiwavelet is developed from multiresolution analysis (MRA). The difference is that multiwavelets have several scaling functions whereas MRA have one scaling function. Multiwavelets offer superior performance for image processing applications compared with scalar wavelets [19]. Multi wavelet offers short support, orthogonality, symmetry, and vanishing moments. A multiwavelet system can provide better reconstruction while preserving length, good performance at the boundaries and a high order of approximation. Each multiwavelet system is a matrix valued multirate filterbank. A multiwavelet filterbank has "taps" that are  $(N \times N)$  matrices. A filter bank is a structure that decomposes a signal into a collection of subsignals [23]. Depending upon the application, the subsignals help to emphasize specific aspect of the original signal or may be easier to work with than that of the original signal. The structure of a classical filter bank is shown in Fig. 1.

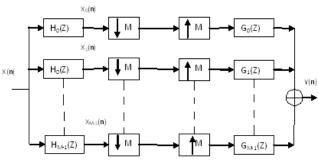


Fig. 1. M-channel filter bank

Perfect Reconstruction Quadrature mirror filters are used to split the input signal into M subbands which are decimated by M in signal decomposition. During reconstruction, M subband signals are decoded, interpolated and recombined using synthesis filters. The Haar Wavelet based M-channel Filter bank (HWF) with M=3 is called the Double Haar Wavelet Transform [24]. The decomposition and Reconstruction filter banks are defined as follows:

$$H = \frac{1}{3} \begin{pmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ 1 & 1 & 1 \end{pmatrix} \qquad G = \begin{pmatrix} 2 & 1 & 1 \\ -1 & 1 & 1 \\ -1 & -2 & 1 \end{pmatrix}$$

Similar to the two dimensional (2-D) orthogonal wavelet transform, the DHWT can also be extended to 2-D signals. Let  $x_0(m, n)$  be an image of  $N \times N$  pixels. The steps of the 2-D discrete double Haar wavelet transform are defined by the following steps [20].

- Step 1: In the horizontal direction, the original image  $x_0(m,n)$  is filtered by the filters  $H_0(z)$   $H_1(z)$  and  $H_2(z)$  respectively. Three images  $x_{00}'(m,n) x_{01}'(m,n)$  and  $x_{02}'(m,n)$  are produced.
- Step 2: In the vertical direction, the three images  $x'_{00}(m,n) \, x'_{01}(m,n)$  and  $x'_{02}(m,n)$  are filtered by the filters  $H_0(z)$ ,  $H_1(z)$  and  $H_2(z)$  respectively. This gives nine images  $x^n_{0j}(m,n) \, 0 \le j \le 8$ .
- Step 3: Down-sampling the images  $x_{0j}^{"}(m,n)$   $0 \le j \le 8$ , with an interval of three, we obtain nine subimages  $x_{0j}(m,n)$   $0 \le j \le 8$ .
- Step 4: Steps 1 to 3 can be repeated on the subimage  $x_{00}(m, n)$  so as to get the other subimages in the next scale.

In two-dimensional DHWT, each level of decomposition produces nine bands of data. The low pass band can further be decomposed to obtain another level of decomposition. Fig. 2 shows the first level of decomposition.

<i>x</i> <sub>00</sub>	<i>x</i> <sub>01</sub>	<i>x</i> <sub>02</sub>
<i>x</i> <sub>10</sub>	<i>x</i> <sub>11</sub>	<i>x</i> <sub>12</sub>
<i>x</i> <sub>20</sub>	<i>x</i> <sub>21</sub>	$x_{22}$

Fig. 2. 1-level DHWT

# III. COMPLEX WAVELET TRANSFORM AND CDHWT

DWT suffers from oscillations, shift invariance, aliasing and lack of directionality as it is based on real valued oscillating wavelets. However, Fourier transform does not suffer from this problem. Because, it is based on complex valued oscillating sinusoids. Therefore, shortcomings of real valued DWT are overcome by the Complex wavelet transform.

CWT can be obtained from Hilbert transform, oscillating cosine and sine components form a Hilbert transform pair as they are 90° out of phase with each other. Hilbert transform is applied to the data. The real wavelet transform was applied to both the original data and the Hilbert transformed data and the coefficients of each wavelet transform were combined to obtain a CWT. Ideal Hilbert transform in conjunction with the wavelet transform effectively increased the support of the wavelets.

One effective approach for implementing an analytic wavelet transform, first introduced by Kingsbury in 1998, was called the Dual-Tree CWT. The Dual Tree CWT employed two real DWTs [2]. The first DWT gives the real part of the transform while the second DWT gives the imaginary part. The analysis and synthesis FBs used to implement the dual-tree CWT and its inverse was illustrated in Figs. 3 and 4.

In this paper, we present a Complex Double Haar Wavelet Transform that combines the advantages of both CWT and M-Channel filter bank. Similar to Dual tree complex wavelet transform CDHWT is also implemented with dual DHWT.  $H_0(n)$ ,  $H_1(n)$  and  $H_2(n)$  are the analysis filter banks gives the real part of DHWT.  $G_0(n)$ ,  $G_1(n)$  and  $G_2(n)$  are the analysis filter banks gives the imaginary part of DHWT. The analysis and synthesis filter banks used to implement the dual-tree CDHWT and its inverse are illustrated in Figs. 5 and 6.

# H<sub>o</sub>(n) H<sub>0</sub>(n) $H_1(n)$ $H_1(n)$ $H_{\mathbf{1}}(n)$ G<sub>o</sub>(n) $G_0(n)$ $G_1(n)$ $G_1(n)$ $G_{\mathbf{1}}(n)$

Fig. 3. Analysis filter bank for the dual-tree discrete CWT

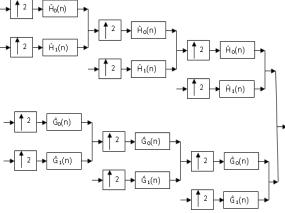


Fig. 4. Synthesis filter bank for the dual-tree discrete CWT

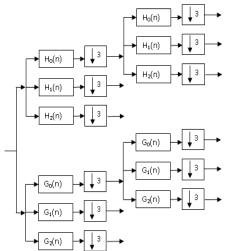


Fig. 5. Analysis filter bank for the CDHWT

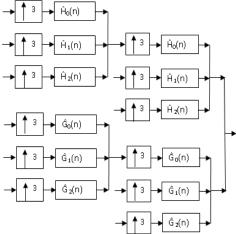


Fig.6. Synthesis filter bank for the CDHWT

#### IV. PROPOSED SCHEME

# A. Embedding Algorithm

A single level CDHWT is applied on the original image X (i,j) and binary watermark image. Singular value decomposition [25],[26] is taken on the subbands  $X_{11}$  of the subband and  $X_{11w}$  of the binary watermark to obtain their Eigen values  $(\sigma, \sigma_w)$ . The Eigen values  $\sigma$  of the host image is replaced by the eigen values  $\sigma_w$  of the watermark image after multiplying with the proper strength factor. Then inverse SVD and Complex DWT were applied to obtain the watermarked image  $X(i, j)^*$ .

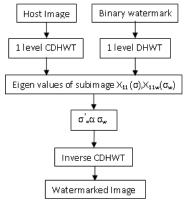


Fig. 7. Watermark embedding algorithm

#### B. Extraction Algorithm

In this paper, we proposed the blind watermarking scheme. Original image and Original watermark are not required to extract the secret message. A one level CDHWT is applied on the test image and their Eigen values of selected sub image are obtained by using SVD. Eigen values are divide with the strength factor in order to extract the binary watermark. Then the extracted watermark is compared with the original watermark, to check whether the test image should contain the watermark or not by measuring the normalized correlation between them.

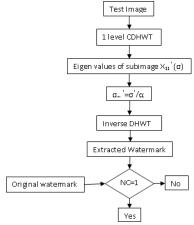


Fig. 8. Watermark extraction algorithm

# V. EXPERIMENTAL RESULTS

The experiments were performed on different gray scale images such as Lena, Baboon, Boat, Fruits, Circles, Rose, and Girl etc. Binary watermark image is of size  $33 \times 33$ . A 1-level CDHWT is applied on the Lena image. A mid frequency band  $(X_{11}s)$  is selected to embed the watermark. A watermarked Lena image is having PSNR value of 50.9419 with no perceptibility problem on watermarked image when using  $\alpha$  at 275 (Lena image). Fig. 9 shows the cover image, original watermark, watermarked image and the extracted watermark.

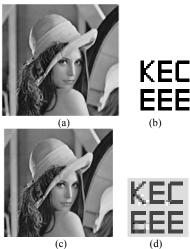


Fig. 9. (a) Cover image; (b) Original Watermark; (c) Watermarked Image; (d) Extracted Watermark

Any watermarking system should be robust against various image processing attacks. It should not be removable by unauthorized users and it should not degrade the quality of the images. There are many attacks against which image watermarking system could be judged. The attacks include JPEG compression, average filtering, rotation, median filtering, Salt and Pepper noise, Gaussian noise, speckle noise and so on. These attacks are applied to the watermarked images to evaluate recovery process. Mean Square Error (MSE), PSNR (Peak Signal to Noise Ratio) and NC (Normalized Cross-Correlation) are used to estimate the quality of extracted watermark. MSE, PSNR and NC [7, 9] are defined as following,

where MSE is defined as follows,

$$MSE = \frac{1}{mn} \sum_{X=1}^{m} \sum_{Y=1}^{n} (f(x, y) - \hat{f}(x, y))^{2} \dots \dots \dots \dots (2)$$

where m and n are size of images, and f(x, y) and  $\hat{f}(x, y)$  are value at (x, y) location of the host and watermarked image,

where and  $p_{ij}$  and  $p_{ij}^*$  are pixel values at  $(i,j)^{th}$  location of the original and recovered watermark patterns respectively.

TABLE I: PSNR VALUES OF WATERMARKED IMAGE

TABLE 1. I SINK VALUES OF WATERMARKED IMAGE							
Image Type	PSNR in db	Normalized Correlation					
Lena	50.9419	0.9961					
Baboon	37.2311	0.9958					
Boat	40.1335	0.9984					
Fruits	39.1722	0.9975					
Rose	41.0753	0.9984					
Girl	39.5898	0.9991					
Moon	50.2254	0.9977					
Cameraman	43.3518	0.9943					
Rohith	44.3346	0.9995					
Circuit	31.985	0.9382					
Circles	35.658	0.98					

PSNR values obtained for various gray scale images are shown in the Table I.

#### A. Robustness to Noise

Robustness to noise is very important in watermarking algorithms. Four kinds of noises were tested. Zero mean Gaussian noise with variance 100, 1% salt and pepper noise, Poisson and speckle noise.

TABLE II: NC UNDER VARIOUS NOISE CONDITION

Image Type	Addition of Noise							
	Gaussian	Salt&Pepper	Poisson	Speckle				
Lena	0.9904	0.9905	0.9903	0.9905				
Baboon	0.9904	0.9904	0.9903	0.9904				
Boat	0.9904	0.9904	0.9903	0.9904				
Fruits	0.9904	0.9904	0.9903	0.9904				
Rose	0.9904	0.9905	0.9903	0.9904				
Girl	0.9905	0.9905	0.9905	0.9905				
Moon	0.991	0.9911	0.9911	0.991				
Cameraman	0.9904	0.9905	0.9903	0.9905				
Rohith	0.9904	0.9905	0.9903	0.9905				
Circuit	0.9906	0.9906	0.9906	0.9906				
Circles	0.9911	0.9911	0.9911	0.9911				

The simulated results of the Normalized Correlation under Various noises are shown in Table II. Fig.10 demonstrates that this algorithm is robust to noise.

Normalised Correlation under various Noise Condition

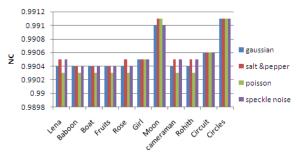


Fig. 10. NC under various Noise Conditions

# B. Robustness to Image Processing Attacks

The watermarking algorithm is also robust to image processing techniques. The popular image processing attacks are histogram equalization, JPEG compression and filtering. The correlation computed from histogram equalized images is shown in Table III. From the results shown in Fig.11 the proposed algorithm is robust to histogram equalization.

TABLE III: NC UNDER HISTOGRAM EQUALIZATION

Image Type	Histogram Equalization
Lena	0.9903
Baboon	0.9903
Boat	0.9904
Fruits	0.9904
Rose	0.9904
Girl	0.9905
Moon	0.991
Cameraman	0.9904
Rohith	0.9904
Circuit	0.9906
Circles	0.991

Watermarked image has been compressed using JPEG

compression with different quality factor as shown in Table IV. A range of QF is typically 1 to 100. As demonstrated in Fig.12 the proposed method is highly robust against JPEG compression with different quality factor in between 1 to 100

TABLE IV: NC UNDER JPEG COMPRESSION

Imaga Tyma	Addition of Noise						
Image Type	1	3	5	10	20		
Lena	0.9903	0.9903	0.9903	0.9903	0.9903		
Baboon	0.9946	0.9944	0.9944	0.9944	0.9944		
Boat	0.9946	0.9944	0.9944	0.9944	0.9944		
Fruits	0.9946	0.9944	0.9944	0.9944	0.9944		
Rose	0.9935	0.9934	0.9934	0.9934	0.9934		
Girl	0.9934	0.9935	0.9935	0.9935	0.9935		
Moon	0.838	0.8381	0.8381	0.8381	0.8381		
Cameraman	0.9486	0.9484	0.9484	0.9484	0.9484		
Rohith	0.9698	0.9697	0.9697	0.9697	0.9697		
Circuit	0.9363	0.9362	0.9362	0.9362	0.9361		

# Normalised Correlation under JPEG Compression

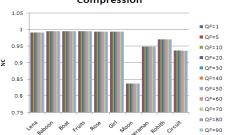


Fig. 12. NC under JPEG Compression

Another popular image processing tool is filter. Two types of filters are tested. Low pass filter and Median filter, which can be considered as case of pixel permutation. The simulated results of the Normalized Correlation for the above mentioned two filtering Conditions are shown in the Table V and Table VI.

TABLE V: NC UNDER AVERAGE FILTERING

Imaga Tyma			Average	Filtering		
Image Type	3×3	5×5	7×7	9×9	11×111	15×15
Lena	0.9939	0.9925	0.9918	0.9915	0.9912	0.9910
Baboon	0.9933	0.9917	0.9911	0.9909	0.9908	0.9906
Boat	0.993	0.992	0.9914	0.991	0.9908	0.9909
Fruits	0.9932	0.9919	0.9913	0.991	0.9909	0.9907
Rose	0.9931	0.9921	0.9915	0.9912	0.991	0.9908
Girl	0.9932	0.992	0.9915	0.9912	0.991	0.9907
Moon	0.9933	0.9923	0.9918	0.9915	0.9912	0.9909
Cameraman	0.9937	0.9926	0.9919	0.9915	0.9912	0.9909
Rohith	0.9923	0.9914	0.9911	0.9909	0.9908	0.9907
Circuit	0.9888	0.994	0.9931	0.9923	0.9917	0.9911
Circles	0.9946	0.9946	0.9938	0.993	0.9925	0.9917

TABLE VI: NC UNDER MEDIAN FILTERING

			Median	Filtering		
Image Type	3x3	5x5	7x7	9x9	11x11	15x15
Lena	0.9942	0.9944	0.9939	0.995	0.9934	0.9948
Baboon	0.9936	0.994	0.9925	0.9942	0.9939	0.9949
Boat	0.9939	0.9946	0.9933	0.9954	0.9944	0.996
Fruits	0.994	0.9944	0.9937	0.9941	0.9929	0.9939
Rose	0.9942	0.9939	0.9944	0.9937	0.9934	0.9934
Girl	0.9939	0.994	0.9929	0.9944	0.9945	0.9958
Moon	0.9938	0.9936	0.9936	0.9935	0.9933	0.9934
Cameraman	0.9941	0.9949	0.9938	0.9952	0.9939	0.9954
Rohith	0.9946	0.9948	0.9939	0.9963	0.9946	0.9968
Circuit	0.9767	0.9921	0.9958	0.9967	0.9971	0.9982
Circles	0.9882	0.9879	0.9899	0.9889	0.9896	0.9902

Figs. 13 and 14 show the experiment results of various gray scale images under filter attacks. We also see that this scheme can resist filter attacks under different window size.

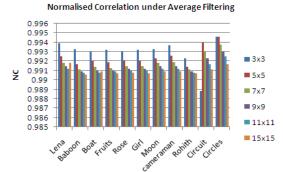


Fig. 13. NC under Average filtering

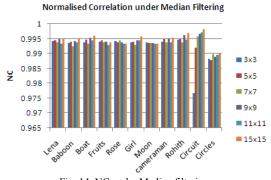


Fig. 14. NC under Median filtering

# C. Robustness to Geometric attacks

Robust against Digital watermarking to geometric attacks is a difficult one that constrains the practical value of watermarking technique. Geometric attacks include rotation, cropping and scaling etc. The correlations computed for various gray scale images under cropping attack are shown in Table VII.

TABLE VII: NC UNDER CROPPING OF AN IMAGE

Image Type	Cropping
Lena	0.9921
Baboon	0.9948
Boat	0.995
Fruits	0.9954
Rose	0.9954
Girl	0.9959
Moon	0.9939
Cameraman	0.9949
Rohith	0.9921
Circuit	0.996
Circles	0.9947

TABLE VIII: NC under Various Angles of Rotation

Image Type	15°	30°	45°	60°	90.	180*	270*		
Moon	0.9935	0.9932	0.9931	0.993	0.9921	0.992	0.992		
Lena	0.9941	0.9934	0.9933	0.9934	0.9884	0.989	0.989		
Rohith	0.9944	0.9938	0.9934	0.994	0.9937	0.9941	0.9939		
Rose	0.9946	0.9942	0.9936	0.9939	0.9929	0.9926	0.9927		
Cameraman	0.9944	0.9946	0.9951	0.9946	0.9935	0.9936	0.9943		
Girl	0.9947	0.9953	0.9951	0.9951	0.9949	0.9948	0.9947		
Boat	0.9937	0.9937	0.9937	0.9936	0.9932	0.9932	0.9932		
Fruits	0.9931	0.9914	0.9904	0.9906	0.989	0.9888	0.9892		
Baboon	0.9937	0.9935	0.9935	0.9936	0.9943	0.9942	0.9961		
Circles	0.9514	0.9633	0.9635	0.9564	0.893	0.8727	0.8799		
Circuit	0.9736	0.9648	0.9548	0.9652	0.9422	0.9444	0.9552		

The simulated results of the Normalized Correlation under various angles of rotation are shown in Table VIII. One can note that this scheme can resist rotational attacks under various angles of rotation.

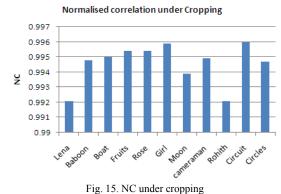
From the results shown in Figs. 15 and 16 the proposed algorithm is robust to geometric attacks.

Table IX shows the image results of extracted watermark

from the Lena image under various attacks. The proposed watermarking scheme is compared with existing recently published papers by Byun *et al* [11], Wang *et al* [12], Jiang *et al*. [13], yuan *et al*. [14], Mahmood *et al* [15], Li [16] *et al*, Peng *et al* [17] and Wei *et al* [18] based on lena image, the results are shown in Tables X-XII.

TABLE IX: IMAGE RESULTS OF EXTRACTED WATERMARK UNDER VARIOUS ATTACKS (LENA IMAGE)

	Addition	JPEG Compression	Histogram Equalization		
Gaussian KEC EEE NC=0.9904	Salt and Pepper KEC EEE NC=0.9905	Poisson KEC EEE NC=0.9903	Speckle KEC EEE NC=0.9903	QF=10 KEC EEE NC=0.9903	KEC EEE NC=0.9903
		Low Pas	s Filtering		
3×3 KEC EEE NC=0.9939	5×5 <b>KEC</b> <b>EEE</b> NC=0.9925	7×7 <b>KEC</b> <b>EEE</b> NC=0.9918	9×9 <b>KEC</b> <b>EEE</b> NC=0.9915	11×11 <b>KEC</b> <b>EEE</b> NC=0.9912	15×15 <b>KEC</b> <b>EEE</b> NC=0.9910
		Median	Filtering		
3×3 <b>KEC</b> <b>EEE</b> NC=0.9942	5×5 <b>KEC</b> <b>EEE</b> NC=0.9944	7×7 <b>KEC</b> <b>EEE</b> NC=0.9939	9×9 <b>KEC</b> <b>EEE</b> NC=0.995	11×11 <b>KEC</b> <b>EEE</b> NC=0.9934	15×15 <b>KEC</b> <b>EEE</b> NC=0.9948
		Rot	tation		
α=0.25 <b>ΚΕ</b> Γ <b>ΕΕΕ</b> NC=0.9942	α=0.5 <b>KEC</b> <b>EEE</b> NC=0.9939	α=0.75 <b>ΚΕ</b> Γ <b>ΕΕΕ</b> NC=0.9939	α=1 <b>KEC</b> <b>EEE</b> NC=0.9937	α=15 <b>ΚΕ</b> Γ <b>ΕΕΕ</b> NC=0.9936	α=30 <b>KEC EEE</b> NC=0.9935
α=45 <b>KEC</b> <b>EEE</b> NC=0.9933	α=60 <b>KEC</b> <b>EEE</b> NC=0.9934	α=90 <b>KEC</b> <b>EEE</b> NC=0.9938	α=180 <b>KEC</b> <b>EEE</b> NC=0.9936	α=270 <b>KEC</b> <b>EEE</b> NC=0.9936	Cropping KEC EEE NC=0.9921



1 ig. 13. We under cropping

Normalised Correlation under various Angles of Rotation 1.02 0.98 0.96 **15** 0.94 0.92 0.9 0.88 0.86 0.84 0.82 Carnet arrain Rohith CHCUIT **POSE** ,400h **270** Gir

Fig.16. NC under various angles of rotation

TABLE X: COMPARISON OF PSNR VALUE OF WATERMARKED IMAGE IN PROPOSED METHOD AND EXISTING METHODS

Methods	PSNR in db
Byun et al	41.95
Wang et al	38.2
Jiang et al.	40.263
Mahmood et al	45.1
Li et al,	40.6
Peng et al	38.764
Wei et al	44.25
Proposed method	50.9419

TABLE XI: COMPARISON OF PROPOSED METHOD AND EXISTING METHODS

UNDER IMAGE PROCESSING ATTACKS

(a) JPEG COMPRESSION								
JPEG	JPEG Mahmood Byun et Wang Lein and Wei Propose							
Compression	et al	al.	et al.	Lin	et al	method		
QF=30%		0.7654	0.15		0.87	0.9903		
QF=40%		0.8148	0.23	0.828	0.95	0.9903		
QF=50%		0.8333	0.26	0.916	0.98	0.9903		
QF = 70%	0.81	0.9383	0.57	0.928	1	0.9903		
QF = 80%	0.89	0.9691		0.945	1	0.9903		
QF= 90%	0.97	0.9938	1		1	0.9903		

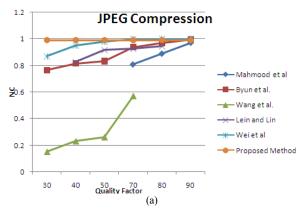
(b) MEDIAN FILTERING							
Median	Li et	Wang	Lein and	Mahmood	Jiang	Wei	Proposed
Filter	al.	et al	Lin	et al	et al	et al	method
3x3	0.35	0.51	0.89	0.92	0.9962	0.88	0.9942

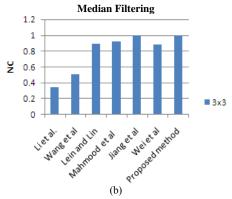
#### (c) Addition of Gaussian Noise

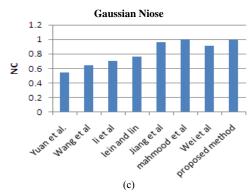
	et al.	et al	al	lin	et al	Mah mood <i>et al</i>		1
Gaussian Noise	0.546	0.64	0.7	0.768	0.9596	1	0.91	0.9904

(d) HISTOGRAM	EOUALIZATION
---------------	--------------

	Yuan et al	Lian and Lin	Wei et al	Proposed Method
Histogram Equalization	0.616	0.935	0.77	0.9903







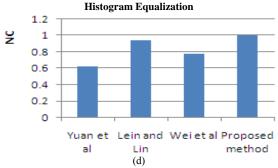


Fig. 17. (a) JPEG Compression; (b) Median Filtering; (c) Gaussian Noise; (d) Histogram Equalization

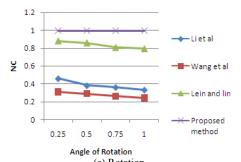
TABLE XII: COMPARISON OF PROPOSED METHOD AND EXISTING METHODS UNDER GEOMETRIC ATTACKS

(0)	ROTAT	TACAL

Rotation	Li et al	Wang et al	Lein and lin	Proposed method
0.25	0.46	0.31	0.88	0.9942
0.5	0.38	0.29	0.859	0.9939
0.75	0.36	0.26	0.808	0.9939
1	0.33	0.24	0.794	0.9937

#### (b) Cropping

	Li et	Jiang et	lein and	Yuan et	Wei et	Proposed
	al	al	lin	al.	al	Method
Cropping	0.61	0.6784	0.88	0.943	0.7	0.9921



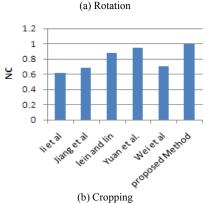


Fig. 18. (a)Rotation; (b)Cropping

From Figs.17 and 18, we can see the PSNR of the watermarked image and the robustness of watermark are far better than those existing methods. The authors claim their method can effectively resist image processing attacks like JPEG compression, median filtering, histogram equalization, addition of noise and geometric attacks like rotation, cropping and can obtain a higher PSNR of the watermarked image.

# VI. CONCLUSION

In this paper, a blind watermarking scheme based on Complex Double Haar Wavelet Transform was proposed. This included the advantages of both multi wavelet and complex representation of an image. Eigen values of the selected sub band are modified by the Eigen values of selected sub band of the binary watermark. During extraction process, watermark was extracted with the help of strength factor there was no need of the original host and watermark for extraction. Experimental results show that the proposed algorithm produced very high imperceptibility and very high robustness against various kinds of attacks like JPEG compression, histogram equalization, low pass filtering,

median filtering, rotation, cropping and addition of noise like Gaussian, Salt& Pepper, Speckle and Poisson noise. From the results of comparison with existing methods, we can believe that the proposed watermarking scheme achieves better imperceptibility and robustness in the watermarking world.

#### REFERENCES

- [1] N. Nikolaidis, and I. Pitas, "Robust image watermarking in the spatial domain", *International journal of signal processing*, vol.66, no. 3, pp. 385-403, May 1998.
- [2] Lei-Da and Bao-Long Guo, "Localised image watermarking in spatial domain resistant to geometric attacks", *International journal of Electronics and communication*, vol.63, no.2, pp. 123-131, February 2009
- [3] Chia-Chin Lin, Wei Liang Tai and Chin Chan Chang, "Multilevel reversible data hiding based on histogram modification of difference images", *International journal of Pattern Recognition*, vol.41, no.12, pp.3083-3096, December 2008.
- [4] Juan R. Hernandez, Martin Amado and Fernando Perez-Gonazalez, "DCT domain watermarking techniques for still images:Detector performance analysis and a new structure", *IEEE transactions on image processing*, vol.9, pp. 55-68, Jan 2000.
- [5] Vassilios Solachidis and Ioannis Pitas, "Circularly symmetric watermark embedding in 2-D DFT", *IEEE transactions on image* processing, vol.10, no.11, pp.1741-1753, Nov 2001.
- [6] Yiwei Wang, John F. Doherty and Robert E. Van Dyck, "A wavelet based watermarking algorithm for ownership verification of digital images", *IEEE transactions on image processing*, vol.11, no.2, pp.77-88, Feb. 2002.
- [7] Chirawat Temi, Somsak Choomchuay, and Attasit Lasakul, "A Robust Image Watermarking Using Multiresolution Analysis of Wavelet", Proc. of International symposium on communication and information technologies, vol.2, pp. 623-62, 2005.
- [8] Shih Hao Wang and Yuan-Pei Lin, "Wavelet tree quantization for copyright protection watermarking", *IEEE transactions on image* processing, vol.13, no.2, pp. 126-144, Feb 2004.
- [9] Kamaran Hameed, Adeel Mumtaz and S. A. M Gilani, "Digital image watermarking in the wavelet transform domain", Proceeding of World Academy of Science, Engineering and Technology(WASET), no.13, May 2006.
- [10] Evelyn Brannock, Michael Weeks and Robert Harrison, "The Effect of Wavelet Families on Watermarking", *Journal of Computers*, vol. 4, no. 6, pp. 554-566, June 2009.
- [11] K. Byun, S. Lee, and H. Kim, "A watermarking method using quantization and statistical characteristics of wavelet transform," *Proc. IEEE, PDCAT*, pp. 689–693, 2005.
- [12] S. H. Wang and Y. P. Lin, "Wavelet tree quantization for copyright protection watermarking," *IEEE Tran. Image Process.*, vol. 13, no. 2, pp. 154–165, Feb. 2004.
- [13] M. Jiang, G. Xu, and D. Yuan, "A novel blind watermarking algorithm based on multiband wavelet transform," *Proc. IEEE ICSP*, pp.857–860, 2004.
- [14] Y. Yuan, H. Decai, and L. Duanyang, "An integer wavelet based multiple logo-watermarking scheme," *Proc. IEEE IMSCCS*, pp. 175–179, 2006.
- [15] A. K. Mahmood and A. Selin, "Spatially Adaptive Wavelet Thresholding for Image Watermarking," in *Proc. IEEE ICME*, Toronto, ON, Canada, pp. 1597–1600, 2006.
- [16] E. Li, H. Liang, and X. Niu, "Blind Image watermarking scheme based on wavelet tree quantization robust to geometric attacks," *Proc. IEEE WCICA*, pp. 10256–10260, 2006.
- [17] B. K. Lien and W. H. Lin, "A watermarking method based on maximum distance wavelet tree quantization," in *Proc. 19th Conf. Computer ,Vision, Graphics and Image Processing*, 2006, pp. 269–276.
- [18] Wei-Hung Lin, Shi-Jinn Horng, Tzong-Wann Kao, Pingzhi Fan, Cheng-Ling Lee and Yi Pan, "An Efficient Watermarking Method Based on Significant Difference of Wavelet Coefficient Quantization", IEEE Transactions on Multimedia, vol. 10, No. 5, pp.746-757, August 2008

- [19] S. Mallat, "A Wavelet Tour of Signal Processing", Academic Press, New York, 1999.
- [20] Nick. Kingsbury, "Dual tree complex wavelet transform", IEEE magazine on signal processing, 2005.
- [21] Mabtoul. S, Ibn Elhaj. E and Aboutajdine, "A robust digital image watermarking method using Dual Tree Complex Wavelet Transform", *IEEE Symposium on Computers and Communications (ISCC)*, pp. 1000 – 1004, September 2008.
- [22] S. Maheswari and K. Rameshwaran, "Complex wavelet transform based adaptive multiple watermarking algorithm for images", CiiT International journal of digital image processing, vol.3, no.5, pp.305-308, March 2011.
- [23] T. Nguyen, and P. P. Vaidyananthan, "Structures for M-channel prefect-reconstruction FIR QMF Banks which yield linear-phase analysis filters", *IEEE Trans. Acoust., Speech, Signal Process.*, vol.38, no.3, pp. 433–446, March 1990.
- [24] Xin Wang, "Moving Window-Based Double Haar Wavelet Transform for Image Processing", *IEEE Transactions on image processing*, vol.15, no.9, pp. 2771 – 2779, September 2006.
- [25] Ramakrishna Kakarala and Philip O. Ogunbona, "Signal Analysis Using a Multi resolution Form of the Singular Value Decomposition", IEEE Transactions on image processing, vol.10, pp. 5, May 2001.
- [26] Gaurav Bhatnagar and Balasubramanian Raman, "A new robust reference watermarking scheme based DWT-SVD", Computer standards and interfaces, vol.31, no.5, pp. 1002-1013, September 2009.



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