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 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 Multiple 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.



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 × 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 $x_{0j}^n(m,n) \ 0 \le j \le 8$.
- 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> ₀₀	<i>x</i> ₀₁	<i>x</i> ₀₂			
<i>x</i> ₁₀	<i>x</i> ₁₁	<i>x</i> ₁₂			
<i>x</i> ₂₀	<i>x</i> ₂₁	<i>x</i> 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.



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



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





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 (σ , σ_w). The Eigen values σ of the host image is replaced by the eigen values σ_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 33×33 . 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 α 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,

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.

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 o	f 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.





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.

ГA	ABLE 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						
Image Tyme		Add	lition of Nois	e		
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	





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

I			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	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



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.

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 shows the image results of extracted watermark

ΓABLE IX: Image Results of Extracted Watermark under Various Attacks (Lena Image)						
	Addition of Noise					Histogram Equalization
Gauss KEI EEI NC=0.9	ian E 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	;	5×5	7×7	9×9	11×11	15×15
KEC		KEC EEE	KEC EEE	KEC EEE	KEC EEE	KEC EEE
NC=0.9	9939	NC=0.9925	NC=0.9918	NC=0.9915	NC=0.9912	NC=0.9910
			Median	Filtering		
3×3	;	5×5	7×7	9×9	11×11	15×15
KEI EEI NC=0.9	9942	KEC EEE NC=0.9944	KEC EEE NC=0.9939	KEC EEE NC=0.995	KEC EEE NC=0.9934	KEC EEE NC=0.9948
			Rot	tation		
α=0.2 KEI EEI NC=0.9	25 E E 9942	α=0.5 KEC EEE NC=0.9939	α=0.75 KEC EEE NC=0.9939	α=1 KEC EEE NC=0.9937	α=15 KEC EEE NC=0.9936	α=30 KEC EEE NC=0.9935
α=4 KEI EEI NC=0.9	5 E 9933	α=60 KEC EEE NC=0.9934	α=90 KEC EEE NC=0.9938	α=180 KEC EEE NC=0.9936	α=270 KEC EEE NC=0.9936	Cropping KEC EEE NC=0.9921



Fig. 15. NC under cropping

Normalised Correlation under various Angles of Rotation



Fig.16. NC under various angles of rotation

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

ΓK	OPOSED MIETHOD AND	EXISTING METHO
	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	Mahmood	Byun et	Wang	Lein and	Wei	Proposed	
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								
	Yuan <i>et al</i> .	Wang et al	Li et al	lein and lin	Jiang et al	Mah mood <i>et a</i>	Wei <i>et al</i>	Proposed method
Gaussian Noise	0.546	0.64	0.7	0.768	0.9596	1	0.91	0.9904
(d) HISTOGRAM EQUALIZATION								
	,	Yuan et	al L	ian and L	in We	ei et al P	roposec	Method
Histogra Equalizat	am tion	0.616		0.935	().77	0.9	903



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

(a) ROTATION							
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	



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.

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