

## Analytic Beta-Wavelet Transform-Based Digital Image Watermarking for Secure Transmission

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**Abstract:** The rapid development in the information technology field has introduced digital watermark technologies as a solution to prevent unauthorized copying and redistribution of data. This article introduces a self-embedded image verification and integrity scheme. The images are firstly split into dedicated segments of the same block sizes. Then, different Analytic Beta-Wavelet (ABW) orthogonal filters are utilized for embedding a self-segment watermark for image segment using a predefined method. ABW orthogonal filter coefficients are estimated to improve image reconstruction under different block sizes. We conduct a comparative study comparing the watermarked images using three kinds of ABW filters for block sizes  $64 \times 64$ ,  $128 \times 128$ , and  $256 \times 256$ . We embed the watermark using the ABW-based image watermarking method in the 2-level middle frequency sub-bands of the ABW digital image coefficients. The imperceptibility and robustness of the ABW-based image watermarking method image is evaluated based on the Peak Signal to Noise Ratio (PSNR) and Correlation coefficient values. From the implementation results, we came to know that this ABW-based image watermarking method can withstand many image manipulations compared to other existing methods.

**Keywords:** Watermarking; ABW; median filtering; JPEG compression; PSNR

### 1 Introduction

Data security becomes an essential issue in the digital world with the rapid growth of digital images and Internet technology accompanied by many security threats on the Internet. For this purpose, many image encryption technologies have emerged and proposed. The existing image encryption technologies use a simple structure and a small key space to reduce the time cost and complexity of algorithms and meet the requirements image information security [1]. Image



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encryption requires high fidelity of the original image due to its sensitivity to the image information [2]. Otherwise, people can easily modify the image content because of the fast advancement in image processing techniques. Therefore, we should focus on the capability of the proposed schemes to be more sensitive to detect any modifications [3]. Most existing approaches are based on scrambling or on information theory in the spatial domain [4]. The limit of this domain resides in the vulnerability to statistical attack [5]. A second family technique uses the chaotic systems where the pixel location is changed by using a 2D chaotic map [6]. These techniques suffer significantly from the sensitivity to the initial and external parameters [7]. The frequency-domain encryption algorithms use some common transforms like discrete cosine transform (DCT), discrete Fourier transform (DFT), and discrete wavelet transform (DWT) to transform the images from spatial to frequency domain first and then get the frequency coefficients and change their positions through those specific rules [8]. The frequency-based approach has strong initial value sensitivity and robust against external attacks [9]. In contrast, the wavelet transform has a low computational complexity that can effectively save the computational cost and reduce the loss of image information in the decryption operation. Nowadays, simplifying the access to digital content has facilitated hacking of image, audio, and video, which can be tampered with and copied. The owner of digital products is threatened severely by the tampered copyright. Therefore, digital watermarking technology is suggested as an effective method to protect copyright authentication and product tracking [10,11]. The protections of different digital contents often require different watermarking algorithms [12].

To enhance the adaptive ability of image watermarking, we proposed the utilization of a new wavelet family named Beta Wavelet family for proposing an efficient image watermarking scheme. The main contribution of the proposed algorithm is as follows: It uses Beta wavelet family to tune many parameters related to the Beta function [13] like order and support parameters and the classical parameters of wavelets like dilation and translation to improve coefficients approximation that achieves the specific goals of detection and verification. This work exploits the ABW advantages to design a robust watermarking approach for protection and verification processes.

The rest of the paper is organized as follows: Section 2 presents the mathematical background of the parametric Beta Wavelets proposed in the schemas of protection and verification of the encrypted images. Section 3 exhibits the proposed schemes in both protection and verification steps. The results analysis is presented in Section 4. Section 5 concludes the paper with future trends.

## 2 Related Work

Image protection and verification have been investigated in the literature. Some of cryptosystems and digital watermarking approaches are proposed to help pushing efforts accordingly. The researchers [14] proposed a content-based image verification cryptosystem for insecure communication channels. First, they transferred the original image into frequency domain using DCT, and they used the mid frequency sub-band for watermarking. Second, the marked image is then encrypted using the chaotic baker map for additional security.

The authors in [15] utilized the dual DCT and chaotic baker mapping to obtain an efficient image cryptosystem. First, they applied the DCT transformed on the red, green, and blue planes of the input image, where each plane was handled independently. Second, they shuffled the DCT coefficients of each plane using chaotic baker mapping. Finally, they applied the inverse DCT and merged the three planes to obtain the encrypted image.

Another paper [16] utilized Fractional Fourier Transform (FrFT) and chaotic logistic mapping to produce secure optical encryption scheme. First, it shuffled and modulated each color channel of the original image using 2D logistic map and random mask, respectively. Second, it transformed the resulted shuffled image of the red, blue, green channels using optical-based FrFT. Again, it shuffled and modulated the transformed images using 2D logistic map and another random mask. Finally, it applied the inverse FrFT and obtained the cipher image.

Authors in [17] presented a multilayer protection scheme for medical images. They watermarked the Electronic Patient Record (EPR) in the original image containing essential health information. They compressed and encrypted the EPR using Huffman algorithm and Quantum logistic map, respectively. Then, they obtained hash value using SHA-256 and embedded it into the image for authentication purposes. After that, they applied the Lifting Wavelet Transform (LWT) on the resulting image's planes. The modified EPR and hash value were to be stored into the different LWT frequency bands, resulting in Watermarked Medical Image (WMMI). Each color layer of the WMMI image permuted using Arnold transformation and then compressed using a lossless Huffman scheme. Finally, they utilized the LWT (HH and LH) frequency bands to embed the resulted WMMI.

Another paper [18] proposed a double fragile watermark system based on diffusion watermark and authentication watermark to protect image integrity. Diffusion watermarking is composed of a cover image and two random sequences. Furthermore, the authors used the diffusion operations, 3D Arnold transformation, and DNA coding to confront chosen cover-image attacks. They utilized the resulted watermark image using DNA encoding to obtain the authentication watermark and cover images. For more security, the authors considered the Latin square permutation and double-layered embedding to perform scrambling to the authentication watermark.

The authors in [19] proposed an image verification scheme by exploiting double-random phase encoding (DRPE) incorporating chaotic mapping. Its proposed generated mask is produced using an image-based chaotic Lorenz system instead of a conventional random-based mask. For verification and authentication, after encrypting the original image using the new proposed DRPE, it is compared with the encrypted reference image in a database by calculating the peak-to-correlation energy.

Technical literature [20] presented a new transmission and integrity verification framework. This framework is robust and reliable for verifying the integrity of HEVC frames transmitted through insecure communication channels. Firstly, the transmitted HEVC frames are divided into a number of blocks with specific block size. The article used a discrete transform for watermarks self-embedding among blocks depending on a predefined mechanism. The suggested schema demonstrates the suitability of the proposed transmission framework for different multimedia cybersecurity applications.

In [21], the authors proposed a crypto-based algorithm that provides confidentiality, authenticity, and integrity for the pixel data and the header data of DICOM files to secure medical image exchanges over public networks. They applied strong cryptographic primitives utilizing internally generated security data, such as encryption keys, hashing codes, and digital signatures. They used strong cryptographic primitives, utilizing internally generated encryption keys, such as encryption keys, hashing codes, and digital signatures to provide the required security services.

The authors of [22] proposed a new joint watermarking/encryption algorithm to verify the reliability of medical images in encrypted and spatial domains. Their algorithm combines a substitutive watermarking algorithm, the quantization index modulation (QIM) with a block cipher

algorithm, and an Advanced Encryption Standard (AES). The proposed solution gives access to the image integrity outcomes and their origins, although they are stored encrypted.

### 3 Analytical Beta Wavelets

Beta Wavelet (BW) is a family of wavelets derived from beta function under certain conditions. Unlike other wavelet forms, BW can be generated based on a proper selection of beta function parameters. The beta distribution is given as follows [13]:

$$(x, p, q, x_0, x_1) = \begin{cases} \left(\frac{x-x_0}{x_c-x_0}\right)^p \left(\frac{x_1-x}{x_1-x_c}\right)^q & x \in [x_0, x_1] \\ 0 & \text{otherwise} \end{cases} \quad (1)$$

where  $x, p, q, x_0, x_1 \in \mathbb{R}$ , with  $p, q > 0$ ,  $x_0 < x_1$ , and  $x_c = \frac{px_1+qx_0}{p+q}$

Indeed, the beta function holds some properties as follows:

1. Beta distribution at the boundary of interval  $[x_0, x_1]$  equals zero, that is  $\beta(x_0) = \beta(x_1) = 0$ .
2. Beta distribution at the centroid,  $x_c$ , equals 1, that is  $\beta(x_c) = 1$ .
3. Evaluation of the derivative of the beta function with respect to  $x$  at  $x_0, x_1$  or  $x_c$  equals zero. That is, knowing that:

$$\frac{d\beta(x)}{dx} = \frac{px_1 + qx_0 - (p+q)x}{(x-x_0)(x_1-x)} \beta(x) \quad (2)$$

$$4. \frac{p}{q} = \frac{x_c-x_0}{x_1-x_c}$$

5. The second derivative of the beta function is given as follows:

$$\frac{d^2\beta(x)}{dx^2} = \beta(x) A(x) \quad (3)$$

where

$$A(x) = \frac{1}{(x-x_0)(x_1-x)} \left[ \frac{1}{(x_1-x)} - \frac{1}{(x-x_0)} - (p+q)(x+1) + px_1 + px_0 \right] \quad (4)$$

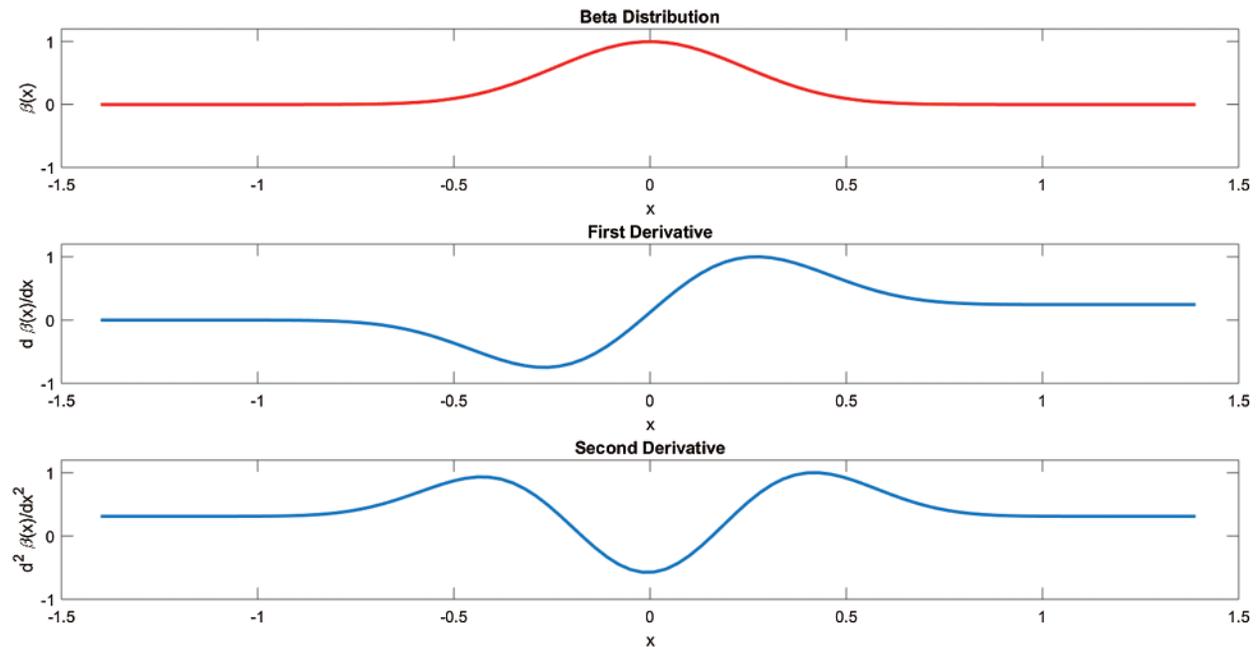
Generally, the  $n$ th derivative of the beta function is given as the following:

$$\begin{aligned} \frac{d^n\beta(x)}{dx^n} &= \left[ (-1)^n \frac{n!p}{(x-x_0)^{n+1}} + \frac{n!q}{(x_1-x)^{n+1}} \right] \beta(x) + P_n(x) P_1(x) \beta(x) \\ &+ \sum_{i=1}^n C_n^i \left[ (-1)^n \frac{(n-i)!p}{(x-x_0)^{n+1-i}} + \frac{(n-i)!q}{(x_1-x)^{n+1-i}} \right] P_1(x) \beta(x) \end{aligned} \quad (5)$$

with  $P_1(x) = \frac{p}{(x-x_0)} - \frac{q}{(x_1-x)}$ , and  $P_n(x) = (-1)^n \frac{n!p}{(x-x_0)^{n+1}} - \frac{n!q}{(x_1-x)^{n+1}}$

The last property is crucial because the beta function's derivative is the essence of beta wavelets. Based on Eq. (5), when ( $n = 0$ ) the result is identical to that one in Eq. (1), which represents beta distribution. However, when ( $n > 0$ ), the beta function takes different wavelet shapes based on derivative order. For instance, Fig. 1 shows the form of beta function in the first row, the second row represents the first derivative of such function, and the third row

depicts the second derivative. Beta function and its generated wavelets achieve three characteristics, localization, oscillation, and admissibility. The proof of these properties is left for the reader in [5].



**Figure 1:** Beta function and its derivatives

Orthogonal multiresolution analysis is a powerful tool to create the basis of orthogonality. Multiresolution can be generated from interpolating scaling functions such as beta wavelet to ease the estimation of wavelet filter coefficients. To illustrate that, assume compact support wavelet in the interval  $[-N/2, N/2]$ , and it is given as:

$$\psi\left(\frac{N}{2}\right) = 2 \sum_{k=-N}^N g_k \phi\left(2\frac{N}{2} - k\right) = 2(g_{-N}\phi(2N) + g_{-N+1}\phi(2N-1) + \dots + g_{N-1}\phi(1) + g_N\phi(0)) \quad (6)$$

Let the interpolation scaling function is defined as follows:

$$\phi(k) = \begin{cases} 1 & k = 0 \\ 0 & k \neq 0 \end{cases} \quad (7)$$

Hence, Eq. (6) becomes

$$\psi\left(\frac{N}{2}\right) = 2g \quad (8)$$

From Eq. (8), filter coefficients  $g_N$ , can be calculated as the following:

$$g_N = \frac{\psi\left(\frac{N}{2}\right)}{2} \quad (9)$$

Generally, the other coefficients that construct filter  $g$  can be computed with the same procedure for each sample as follows:

$$g_k = \frac{\psi\left(\frac{k}{2}\right)}{2} \quad (10)$$

Eventually, the wavelets are orthogonal, and the filter is a quadrature mirror filter if any sample of  $g$  creates an orthogonal base on  $L^2 \in \mathbb{R}$ , that is  $g_n = (-1)^n h_{1-n}$ .

#### 4 The Proposed Image Watermarking Method

In this section, we study the suggested ABW-based image watermarking method in detail. The proposed method is with the potential of image integrity, verification, and tamper detection. It has two modules, watermark embedding module, and verification module. The two modules are explored and explained in detail. The Proposed ABW-based image watermarking method might be considered a fragile self-embedding watermarking process. It depends on segment-based self-watermarking instead of employing external watermarks.

##### 4.1 Protection Module of the Suggested ABW-Based Image Watermarking Method

The protection module of the suggested method is explored stepwise as follows, considering three different Analytic Beta-Wavelet (ABW) orthogonal filters.

- (a) Split the input image,  $f$ , into two equal sub-images,  $f_1$  and  $f_2$ . Again, split the resulting parts into different non-overlapping segments. In our work, we try the block size of  $64 \times 64$ ,  $128 \times 128$ , and  $256 \times 256$ . For example, given the original image of size  $(256 \times 256)$ , when splitting into two equal sub-images ( $f_1$ ,  $f_2$ ), their size will be  $(128 \times 256)$  each. Each sub-image is then divided into  $8 \times 8$  non-overlapped blocks. Therefore, the number of blocks per sub-image is  $(128 \times 256)/(8 \times 8) = 512$ .
- (b) Apply ABW transform to each block in the sub-image  $f_1$  starting from the upper-left corner block.
- (c) Embed the row and column of each block in sub-image  $f_1$  into the row and column of the corresponding transformed block in sub-image  $f_2$ .
- (d) Apply inverse ABW transform to each block in sub-image  $f_1$  starting from the upper-left corner block.
- (e) Apply ABW transform to each block in sub-image  $f_2$ , and repeat steps 3 and 4 for such an image.
- (f) Recompose the resulted two sub-images to obtain the block-based watermarked image.

##### 4.2 The Verification Module of the Proposed Method

The process of verification module can be digested in the following steps:

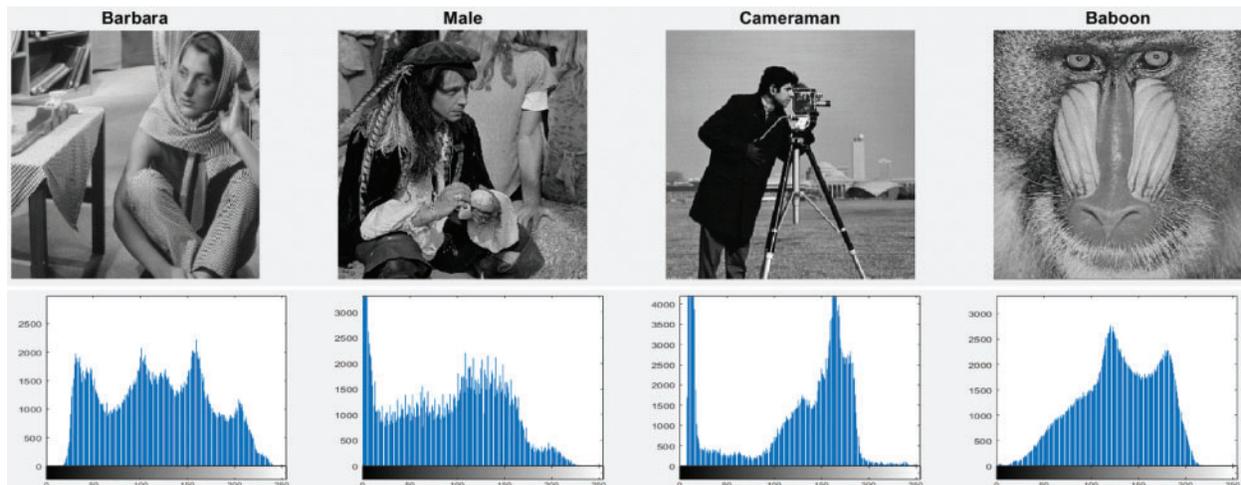
- (a) Receive the watermarked image.

- (b) Split the watermarked image,  $z$ , into two equal sub-images,  $z_1$  and  $z_2$ . Again, split the resulting parts into different non-overlapping segments. In our work, we try the block size of  $64 \times 64$ ,  $128 \times 128$ , and  $256 \times 256$ .
- (c) Apply ABW transform to each block in the sub-image  $z_1$  starting from the upper-left corner block.
- (d) Embed the row and column of each block in the transformed sub-image  $z_1$  into the row and column of the corresponding block in sub-image  $z_2$ .
- (e) Apply inverse ABW transform to each block in sub-image  $z_1$  starting from the upper-left corner block.
- (f) Apply ABW transform to each block in sub-image  $z_2$ , and repeat steps 4 and 5 for such an image.
- (g) Recompose the resulted two sub-images to obtain the extracted image.

## 5 Simulation Results

### 5.1 Experiments Setup and Dataset

We used four standard and widely used images to assess the performance and robustness of the beta wavelets in watermarking and forgery detection. These images are Barbara, Male, Cameraman, and Baboon, as shown in Fig. 2. The four images used in the experiments have different histograms, evident from Fig. 2, which depict the variation of the results on these images.



**Figure 2:** The standard images used in experiments with their histograms

### 5.2 Watermarking Using Beta Wavelets

The watermark is embedded into images by applying the safeguard method to the beta wavelets approximation coefficients at level 2. The non-sampled wavelet decomposition scheme is used in the experiments to preserve the quality of both the watermarked and the retrieved images.

The authors in [13] generated different bi-orthogonal beta wavelet filters and compared them to traditional wavelet families, such as Daubechies and Morlet wavelets. They tested such beta wavelets in an image compression application. Bi-orthogonal wavelets are less phase distortion,

and they support symmetrical and exact reconstruction than orthogonal wavelet filters. Bi-orthogonal wavelets have two sets of scaling filters, and hence wavelet filters. No prior experiment with beta wavelets for image watermark has been carried out in the literature. The performance of those filters with different Filter orders in the image's watermark is checked and compared with two wavelets in this article: Daubechies (db2) and discrete Meyer wavelets.

Consequently, along with the Daubechies (db2) and discrete Meyer (dmey) wavelets, 17 bi-orthogonal beta wavelet filters with different filter orders are generated and applied to the Barbara image with block sizes of  $64 \times 64$ ,  $128 \times 128$ , and  $256 \times 256$ . The results of this study are provided in [Tab. 1](#). The performance of the BWT is assessed using visual inspection under two evaluation metrics: PSNR, and Correlation coefficient. Given two images  $I_1$  and  $I_2$ , the equations of these measures are:

$$\text{PSNR} = 10 \log_{10} \left( \frac{255}{\frac{1}{MN} \sum_{n=1}^N \sum_{m=1}^M (I_1(m, n) - I_2(m, n))^2} \right) \quad (11)$$

$$\text{Cr} = \frac{\sum_n \sum_m (I_1 - \mu_1)(I_2 - \mu_2)}{\sqrt{\sum_n \sum_m (I_1 - \mu_1)^2 \sum_n \sum_m (I_2 - \mu_2)^2}} \quad (12)$$

where  $\mu_1$  and  $\mu_2$  are the mean of  $I_1$  and  $I_2$ , respectively.

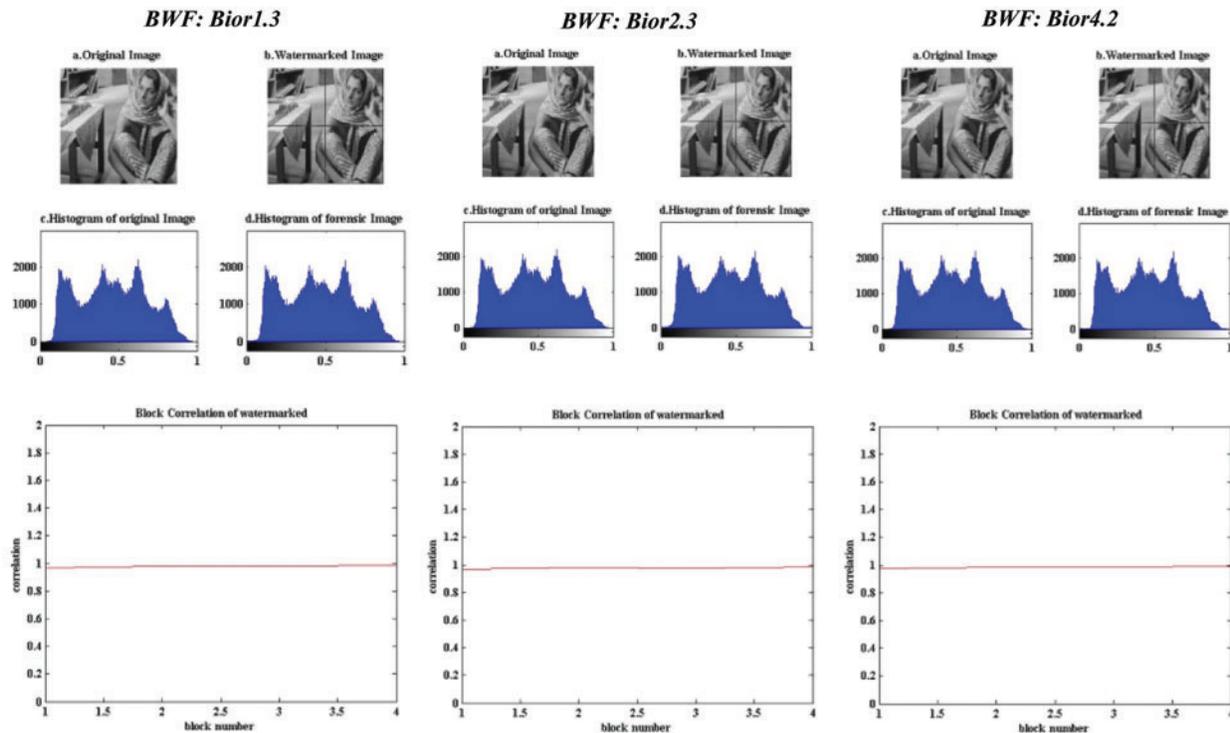
**Table 1:** PSNR and Correlation coefficient values of watermarked images with various Beta wavelet filters for the Barbara image with block sizes  $64 \times 64$ ,  $128 \times 128$ , and  $256 \times 256$

Beta wavelet filter	PSNR			C <sub>r</sub>		
	$64 \times 64$	$128 \times 128$	$256 \times 256$	$64 \times 64$	$128 \times 128$	$256 \times 256$
Bior 1.3	28.0611	33.1331	33.1714	0.9337	0.9793	0.9791
Bior 1.5	28.0603	33.0822	33.1693	0.9339	0.9791	0.9792
Bior 1.7	28.0205	32.6380	33.0464	0.9343	0.9776	0.9790
Bior 2.1	27.9193	33.1370	33.0343	0.9316	0.9793	0.9790
Bior 2.3	27.9184	33.1113	33.0334	0.9317	0.9792	0.9785
Bior 2.4	27.9021	32.8812	32.9961	0.9319	0.9784	0.9785
Bior 2.5	27.9108	31.9343	32.5639	0.9310	0.9744	0.9770
Bior 2.6	27.9108	32.9812	33.0178	0.9318	0.9787	0.9785
Bior 2.7	27.9177	33.0944	33.0323	0.9317	0.9792	0.9785
Bior 3.2	28.5637	31.5690	33.7217	0.9409	0.9713	0.9817
Bior 3.3	28.3537	31.4638	33.6922	0.9409	0.9708	0.9817
Bior 3.4	28.4990	31.0926	33.5275	0.9407	0.9689	0.9813
Bior 4.1	28.2647	32.8723	33.3862	0.9366	0.9780	0.9801
Bior 4.2	29.2301	30.0958	34.4807	0.9487	0.9594	0.9844
Bior 5.3	27.9252	27.1127	33.1375	0.9317	0.9238	0.9790
Bior 5.4	27.9938	27.9990	31.5637	0.9328	0.9366	0.9699
Bior 5.5	28.0853	27.9101	33.2794	0.9340	0.9354	0.9796
DB2	27.7940	33.1176	32.9531	0.9290	0.9792	0.9780
DMEY	27.8467	33.1273	32.9959	0.9300	0.9793	0.9782
Haar	28.0611	33.1331	33.1714	0.9337	0.9793	0.9791

The findings in [Tab. 1](#) show that the beta wavelet filter’s behavior differs depending on the PSNR and Cr values for the same image across the three block sizes. As shown in [Tab. 1](#), the best performance of each filter out of the three block sizes is highlighted with a bolded font. The bior1.3 filter, for example, has better PSNR with  $256 \times 256$  blocks than other block sizes, while its Cr coefficient has the best value with  $128 \times 128$  blocks. For both PSNR and Cr, the bior2.3 filter shows the best results with the same block size. The PSNR and Cr coefficients for the bior4.2 filter, on the other hand, indicate that the block size  $256 \times 256$  produces the best results as compared to other block sizes. This filter, bior4.2, also has the best values of all the beta and common wavelet families tested in this study. Comparing the biorthogonal beta filters in this paper with the wavelets Haar, db2, and dmey allows us to use them in watermarking simulation experiments, as shown in the following section. Due to the difficulty of testing all filters systematically in the following experiments, only three beta filters, bior1.3, bior2.3, and bior4.2, are chosen to represent the remaining beta filters, each of which has two possible performance measure variants.

### 5.3 Evaluation of Watermark Embedding

As mentioned in the previous section, the performance of the three selected beta wavelet filters bior1.2, bior2.3, and bior4.2 is investigated further as the experiments progress. The quality measures of the four images when inserting the watermark using the safeguard procedure with beta wavelets under block size of  $256 \times 256$  is shown in [Fig. 3](#).



**Figure 3:** Protection process visual results of the original and watermarked images with Beta Wavelet for the four tested images of block size  $256 \times 256$

The plots in Fig. 4A show that when changing the block size, the performance of all filters varies. Bior4.2 has the best performance with the  $64 \times 64$  block size, while the other two filters have relatively close MSE measurements. Increasing the block size to  $128 \times 128$  dramatically improves the results of the first two filters, while slightly enhancing the error measure for the bior4.2, except for the Baboon image. The error in the Baboon image increases with  $128 \times 128$  block size compared to the  $64 \times 64$  block size for the bior4.2 filter. Applying the beta wavelet filters to images with different characteristics and different histograms may affect their output; the Baboon image is an example. All filters have far better outcomes with the  $256 \times 256$  block size than other block sizes.

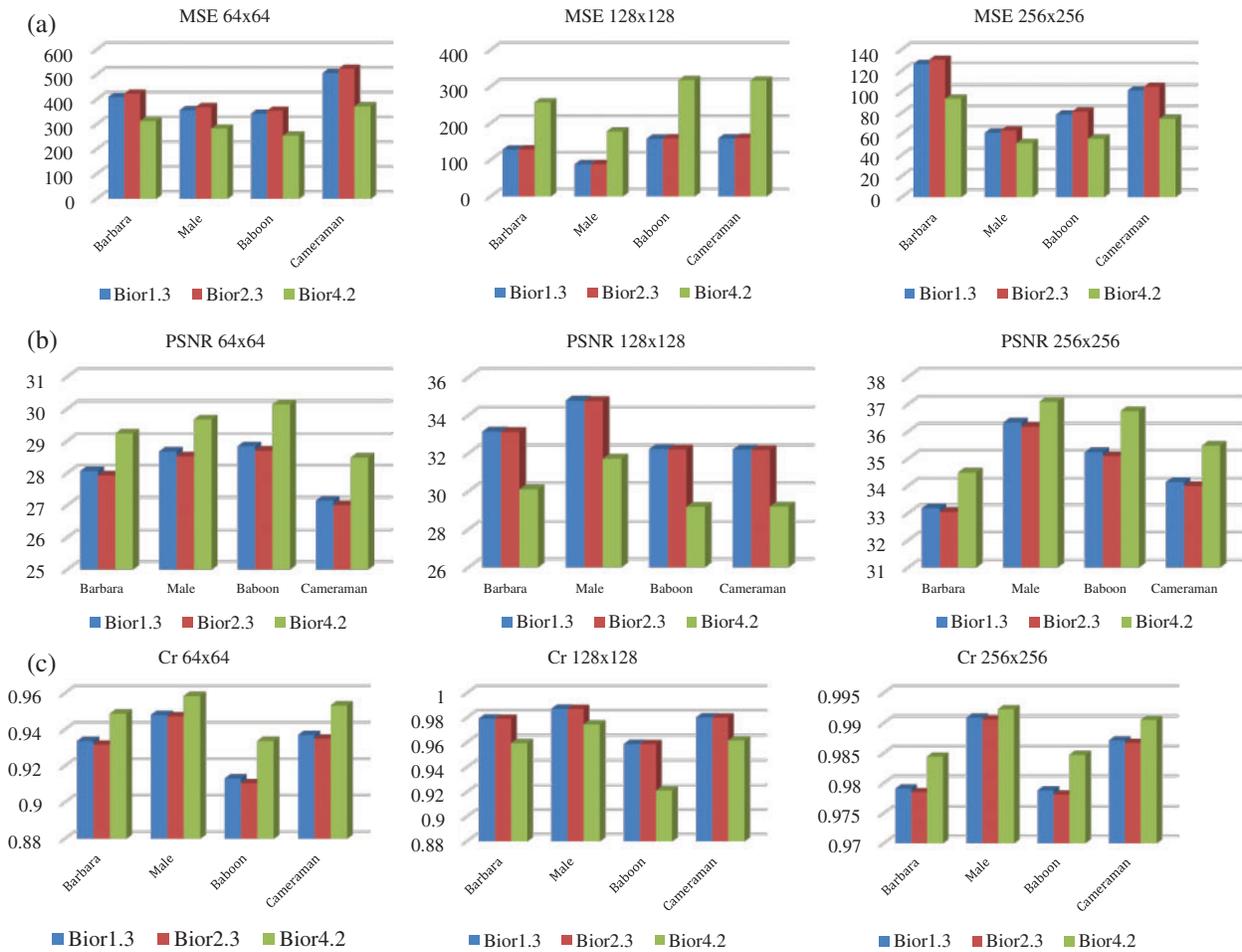
Tab. 1 and Fig. 4A clearly show that the performance measurements of the  $256 \times 256$  block sizes outperform those of the  $64 \times 64$  and  $128 \times 128$  block sizes. Although the bior4.2 beta wavelet with block size  $256 \times 256$  produces the best results, the bior1.3 and bior2.3 filters have comparable results, with the bior1.3 filter slightly outperforming the latter. The same behavior is repeated for the PSNR measure for the three beta wavelet filters as shown in Fig. 4B. The plots in Fig. 5C indicate that the Cr for the Baboon image has less performance compared to the other images used in the experiment for all filters. Although there are a lot of details in the Baboon image, most of these details are sporadic. Therefore, much of the energy for this image is based on the approximation coefficients that are subjected to the watermark at level 2 in this paper. The Baboon image displays less immunity to distortion when we attempt to extract it from the watermarked image. The visual quality of the watermarking process of the three biorthogonal filters and their effect on the tested images are shown in Fig. 3. The associated block-correlation provides the quality change assessment for each block on the images.

In this experiment, the images are extracted using the verification process with the Beta wavelet for block sizes  $256 \times 256$ ,  $128 \times 128$ , and  $64 \times 64$  without any attack. This method aims to ensure that the images are obtained from its watermarked version with minimal degradation. Again, the quality of the extracted images is measured by three metrics: PSNR, and Cr. Also, the extracted images are shown in each of the following Figs. 5–7, next to the original copy to visually assess the effect of the watermarking and extraction processes.

Tab. 2 indicates the performance of the three previously selected beta wavelet filters for all images based on the three different block sizes  $64 \times 64$ ,  $128 \times 128$ , and  $256 \times 256$ . Similar to the outcome of the previous experiments, almost in all cases, performance indicators show that the  $256 \times 256$  block size outperformed the block sizes  $64 \times 64$  and  $128 \times 128$ . Also, the bior4.2 filter has the best results on  $256 \times 256$  block size against all other filters with various block sizes when applied to the four tested images.

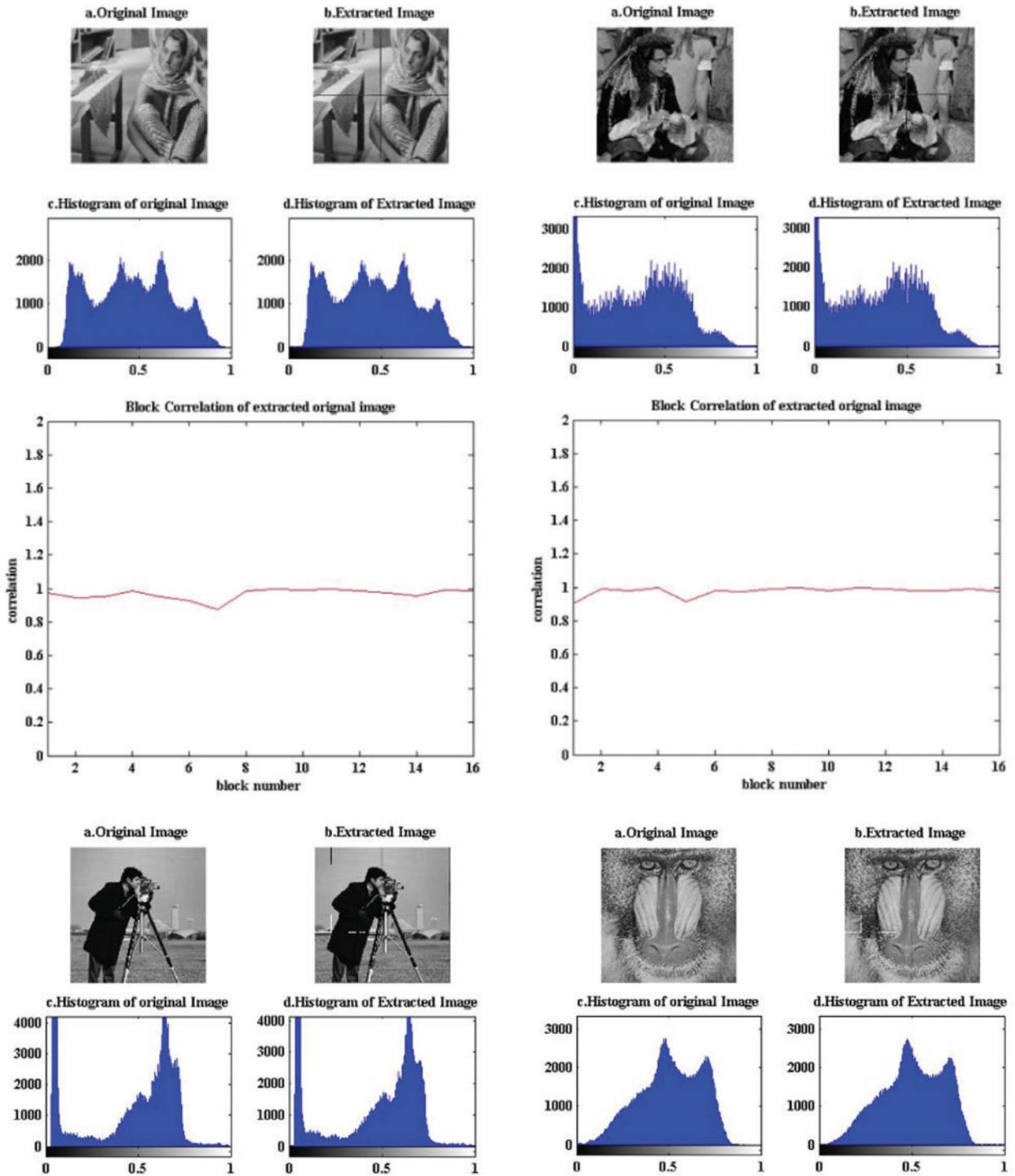
The block correlation plots shown in the Figs. 5–7 next to the extracted images indicate good results for  $256 \times 256$  and  $128 \times 128$  block sizes for all filters performed well as expected. However, block correlation measure is dropped in Cameraman image when dividing the image into a small number of blocks. For the  $64 \times 64$  block size, all correlation coefficients (Cr) have a fluctuating behavior. Cameraman results show the worst case where the Cr alternate between zero to higher

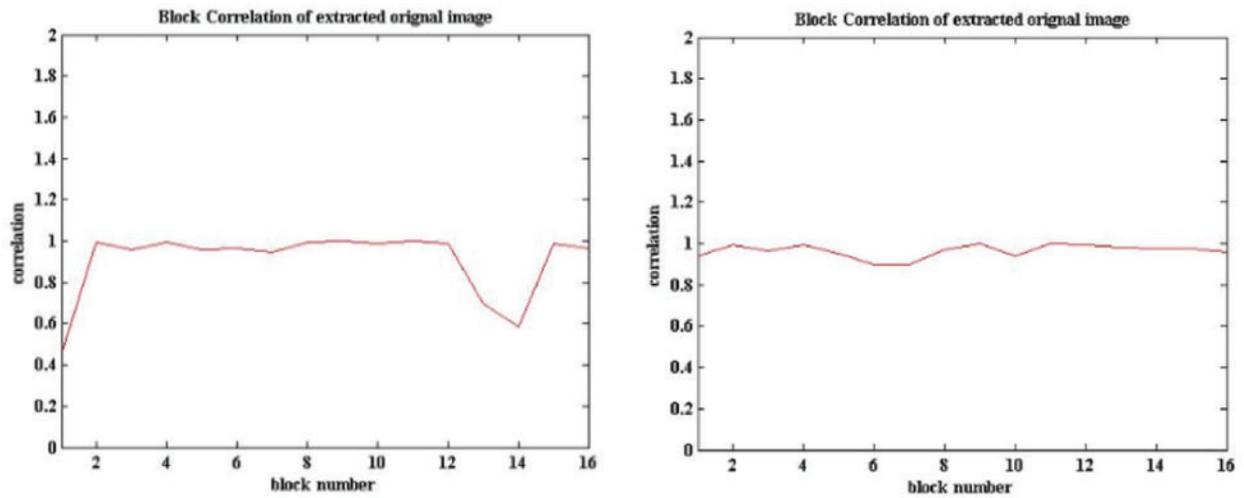
values closer to one. This behavior is due to the large dark areas found in the image and the nonuniform intensity distribution, evident from its histogram, which affects the image extraction. On the contrary, the Baboon image has the best stable readings due to its lack of these large dark areas. The same happens for the Barbara and Male images with most blocks that do not have large dark spots.



**Figure 4:** (A) MSE of watermarked images using the safeguard procedure with beta wavelet filters for block sizes  $64 \times 64$ ,  $128 \times 128$ , and  $256 \times 256$  applied to all images. (B) PSNR of watermarked images using the safeguard procedure with beta wavelet filters for block sizes  $64 \times 64$ ,  $128 \times 128$ , and  $256 \times 256$  applied to all images. (C) Cr of watermarked images using the safeguard procedure with beta wavelet filters for block sizes  $64 \times 64$ ,  $128 \times 128$ , and  $256 \times 256$  applied to all images

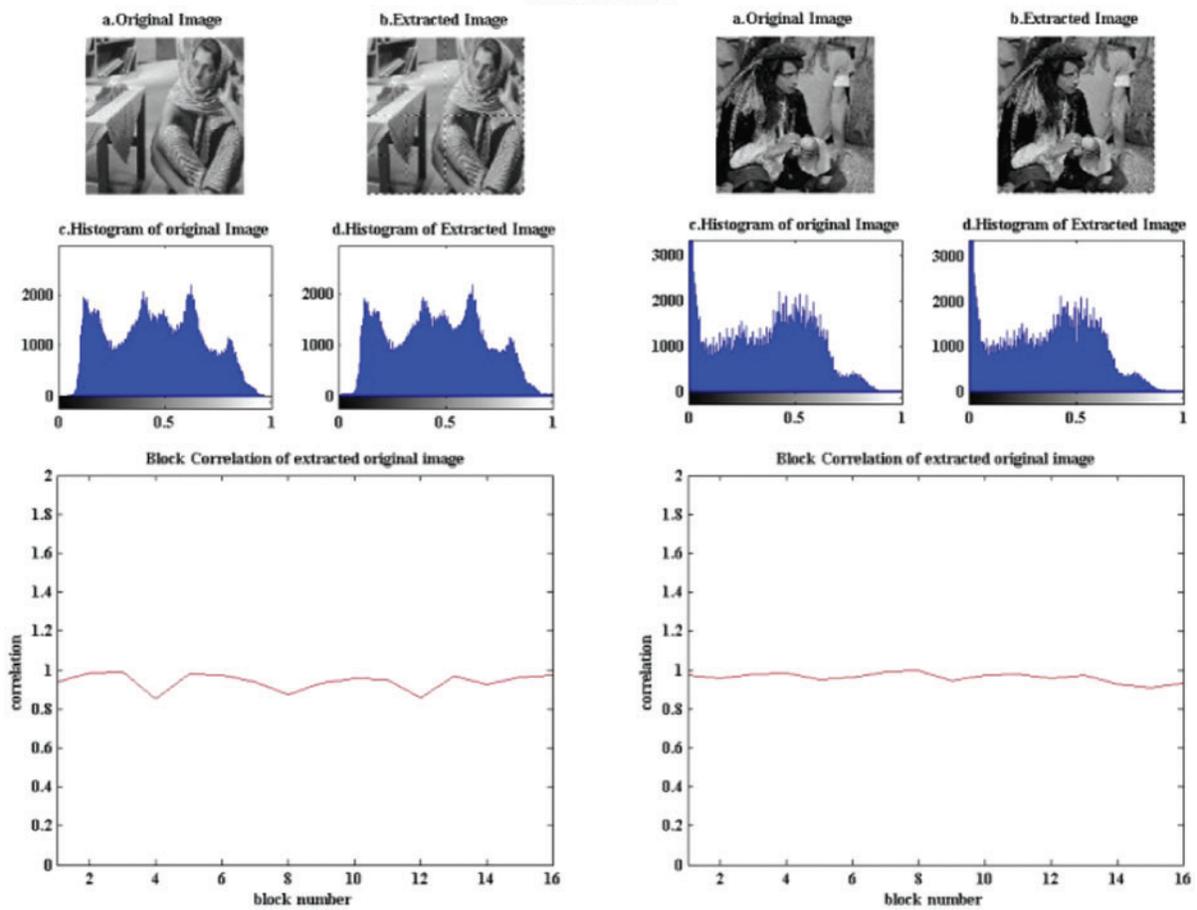
**BWF: Bior4.2**

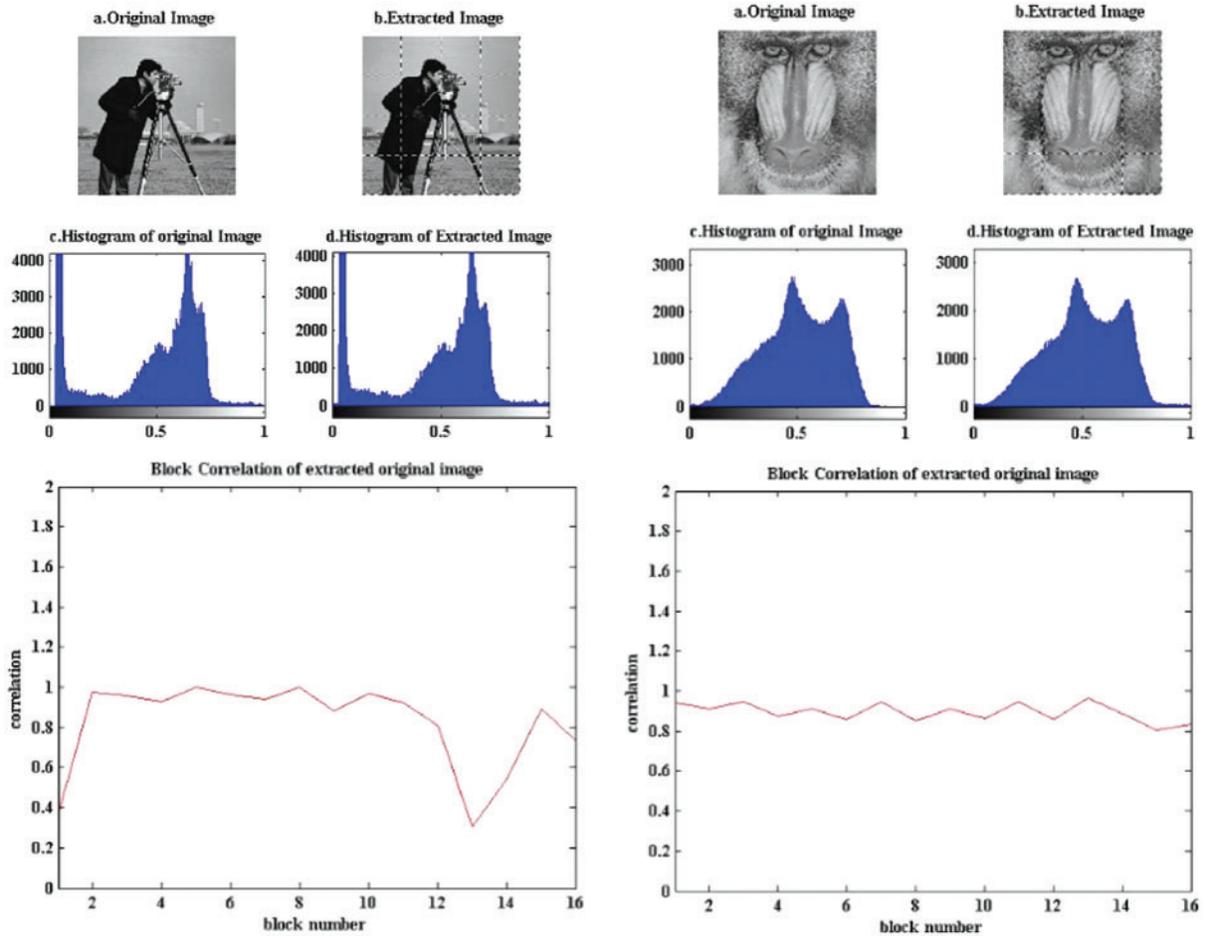




**Figure 5:** Verification process visual results of the original and extracted images with Beta wavelet for the four tested images of block size of  $256 \times 256$

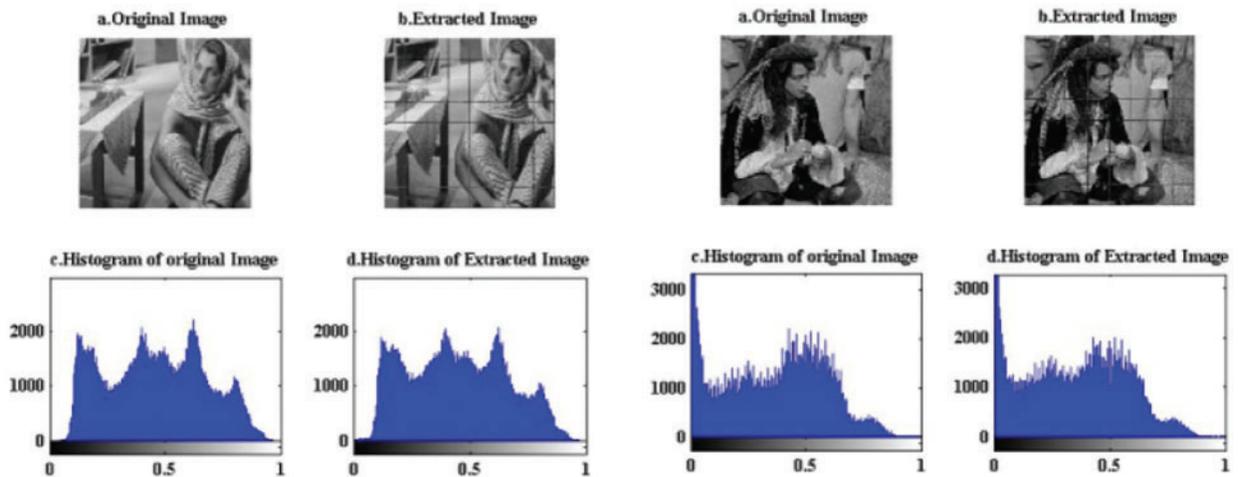
***BWF: Bior4.2***

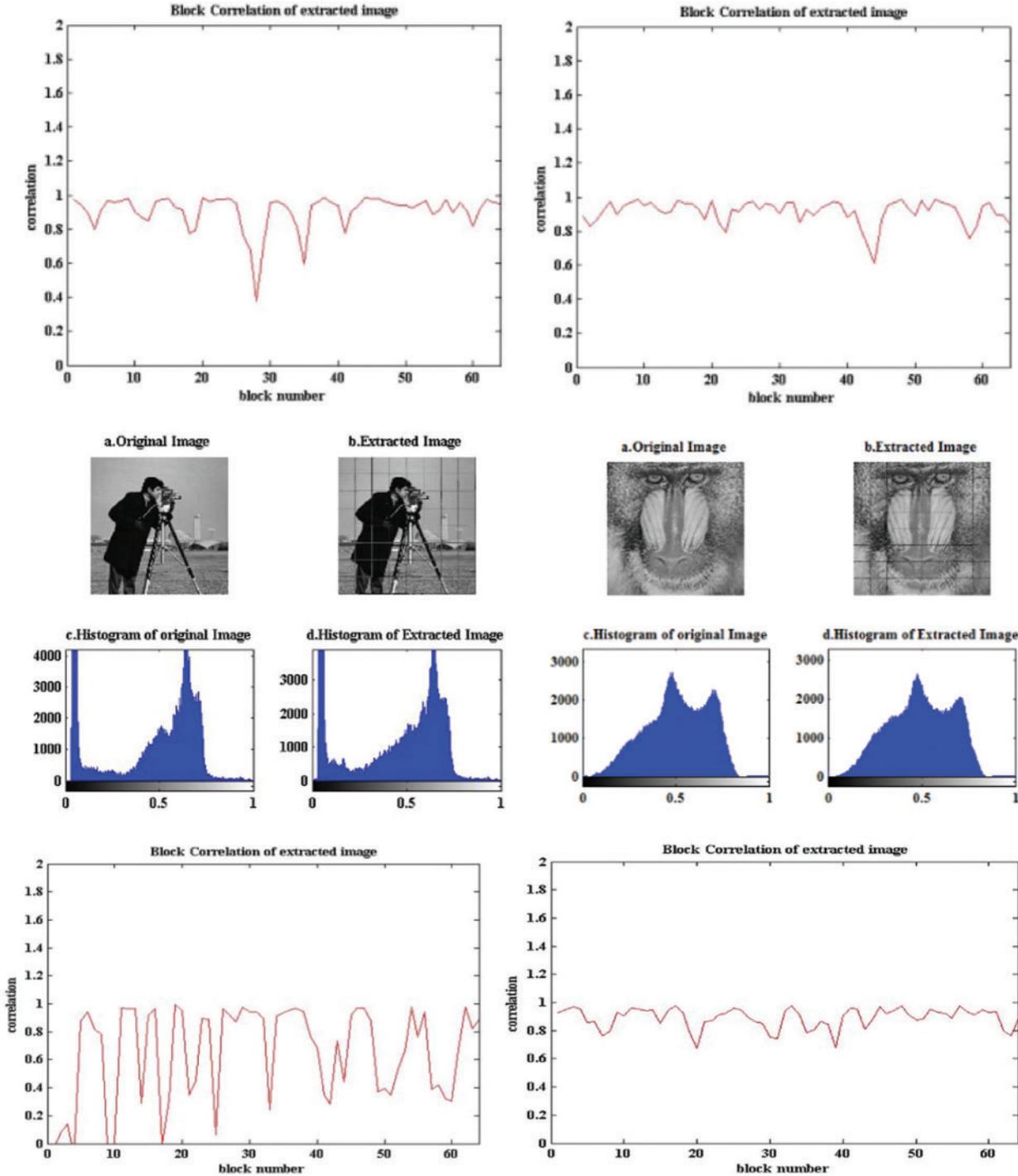




**Figure 6:** Verification process visual results of the original and extracted images with Beta wavelet for the four tested images of block size of  $128 \times 128$

***BWF: Bior4.2***





**Figure 7:** Verification process visual results of the original and extracted images with Beta wavelet for the four tested images of block size of  $64 \times 64$

**Table 2:** PSNR and Correlation coefficient values of extracted images using the verification process with Beta wavelet for the four tested images of various block sizes of  $256 \times 256$ ,  $128 \times 128$ , and  $64 \times 64$

Beta wavelet filter	Block size	PSNR				$C_r$			
		Brabra	Male	baboon	Cameraman	Brabra	Male	baboon	Cameraman
Bior1.3	$256 \times 256$	31.397	33.95	33.492	32.244	0.9701	0.984	0.9684	0.9801
	$128 \times 128$	33.133	34.76	32.219	32.191	0.9793	0.987	0.9587	0.9802
	$64 \times 64$	28.061	28.67	28.834	27.141	0.9337	0.948	0.9131	0.9370
Bio2.3	$256 \times 256$	31.653	33.85	33.389	32.146	0.9695	0.984	0.9678	0.9797
	$128 \times 128$	33.111	34.74	32.192	32.165	0.9792	0.987	0.9586	0.9801
	$64 \times 64$	27.919	28.52	28.688	26.999	0.9317	0.946	0.9105	0.9350
Bior4.2	$256 \times 256$	32.177	33.66	33.956	32.7301	0.9736	0.983	0.9711	0.9822
	$128 \times 128$	30.097	31.71	29.177	29.1922	0.9594	0.975	0.9211	0.9616
	$64 \times 64$	29.232	29.67	30.139	28.4806	0.9487	0.958	0.9337	0.9532

## 6 Conclusions

The paper presented an efficient ABW-based image watermarking protection and verification framework. The proposed ABW-based image integrity verification framework applied a certain ABW orthogonal filter for embedding internal segment-based watermarks into other segments of the transmitted image. Three different ABW orthogonal filters are examined in the proposed ABW-based image integrity verification framework. Simulation tests demonstrated that the possibility of watermark protection and verification using the suggested ABW-based image watermarking framework. Additionally, the proposed method provided a high robustness against multimedia attacks. Also, tampering and forgery detection simulations indicate superior results. The test results also indicated the high sensitivity of the suggested ABW-based image watermarking framework to detect different types of image tampering, although the received tampered image appeared to be visually not manipulated. Finally, we showed that the proposed method can be applied to provide a confidential image communication and detect any forensic operations.

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