

IMAGE AUTHENTICATION USING SUBBAND CODING

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ABSTRACT

In this paper, an image accreditation technique by embedding digital watermark into image is proposed. The proposed method for the digital watermarking is based on the context-based wavelet transform. First, unlike most of previous works which uses a random number of a sequence of bits, and can only be detected by employing the detecting methods. An experimental threshold is chosen and compared to determine whether a sequence of random signal is the watermark. In our approach, we embed the watermark with visual recognizable patterns into the images by modifying the frequency part of the image. In the proposed method, original image data and grey-level digital watermark are decomposed into wavelet coefficients. The performance of the proposed watermark is robust to a variety of signal distortions, such as JPEG, image cropping, sharpening, median filtering, and incorporating attacks.

Index terms—Digital watermark, discrete wavelet transform, context-based wavelet transform, JPEG compression, image processing.

1. INTRODUCTION

Multimedia production and distribution is all digital from the authoring tools of content providers to the reviews. The success of the Internet allows for the wild distribution of multimedia data in an effortless manner. From the viewpoint of media producers and content providers, the possibility for unlimited copying and illegal distribution of privately owned digital data is undesirable because it may cause considerable financial loss. One potential solution for declaring the ownership of the image is the embedding of digital watermarks

into multimedia data. Watermarking is a technique for labeling digital pictures by hiding secret information into the images. The main intended application is to prove ownership of an image for copyright protection. This idea is to embed secret information into the image that can neither be removed nor decoded without the required secret keys. The owner may add a watermark in the multimedia that authenticates the legal copyright holder and that cannot be manipulated or removed without impairing the multimedia data so that they are of no commercial value any more.

Indeed, there are a number of desirable characteristics that a watermark should exhibit. It at least should respect the following requirements for image watermark:

1. **Readability:** A watermark should convey as much information as possible, i.e., the data rate of watermark should be high. A watermark should be statistically undetectable. Moreover, retrieval of the digital watermark can be used to identify the ownership and copyright unambiguously.
2. **Security:** A watermark should in general be secret and must be undetectable by an unauthorized user. A watermark should only be accessible by authorized parties. This requirement is referred to as security of the watermark and is usually achieved by the use of cryptographic keys. As in information security techniques, the details of the digital watermark algorithms must be publishable to everyone. The owner of the intellectual property image is the only one who holds the private secret keys.
3. **Imperceptibility:** One of the main requirements for watermarking is the perceptual transparency. The digital watermark should not be noticeable to the viewer. The embedded watermarking image

should be perceptually invisible. The data embedding process should not introduce any perceptible artifacts into the original image nor should the digital watermark degrade the perceived quality of the image.

4. Robustness: The digital watermark is still present in the image after distortion and can be detected by the watermark detector, especially on the attack from compression. Possible distortions include linear or nonlinear filtering, image enhancements, requantization, resizing, and image compression. The watermark should utilize robust image processing which preserves the desired quality of the image.

In the literature, several digital watermarking algorithms have been proposed with different contributions. Roughly speaking, these contributions can be categorized according to their casting/processing domain, signal type of the watermark and hiding position. A variety of watermarking schemes have been reported in recent years. Such techniques can be broadly classified in two categories: the spatial domain, for example [1-6], and the frequency domain watermarks, for example [7-15]. The earlier watermarking techniques reported were spatial-based approach, the simplest being the ones to modify the least significant bits (LSB) of an image's pixel data [1]. Improvements of these techniques are proposed in [2-6] to be robust against image compression and filtering. Bender *et al.* described a watermarking approach by modifying a statistical property of the image [2]. But such techniques still have relatively low-bit capacity and are not resistant enough to lossy data compression and image processing. For example, a common image cropping operation may eliminate the watermark. Other than spatial-domain-based watermarking, transform-domain-based techniques that can embed a large number of bits of watermark have also been proposed.

In [7], the spread spectrum communication is used in multimedia watermarking. They embed a set of independent and identical distributed sequences drawn from a Gaussian distribution into the perceptually most significant frequency components of the data. Results reported with the *top* (largest in absolute magnitude) thousand coefficients (excluding the DC coefficient) of the DCT of the image show the technique to be

remarkably robust against various image-processing operations. Since this is only a small fraction of the number of significant coefficients in a typical image, there is not much perceptual degradation of the image. C-T Hsu *et al.* embed the watermarks with visually recognizable patterns into the images by selectively modifying the middle-frequency parts of the image. The embedding and extracting methods of the DCT-based approach have been described [9]. Another method, to hide data into the frequency domain to provide extra robustness against attacks is the discrete wavelet transform (DWT) [10-13]. It is known that the wavelet image/video coding, such as embedded zero-tree wavelet (EZW) coding, has potential to play an important role in upcoming image/video compression standards, such as JPEG2000 and MPEG4 due to its excellent performance in compression.

In most of the previous wavelet-based works, for example [10-12], the watermark is a random sequence of bits, and can only be detected by employing the detection theory. Detection involves retrieving the watermark by subtracting the original image from the watermarked image and correlating the retrieved watermark with the original watermark. Therefore, an experimental threshold is chosen and compared to a sharp peak in the cross correlation coefficient to determine whether the image is watermarked. In this paper, we propose a technique for embedding grey-level digital watermarks with visually recognizable patterns into the image. Such kinds of visually recognizable patterns are more intuitive for representing an organization's logo. Moreover, during the verification phase of our work, an extracted visual pattern in conjunction with the similarity measurement will be provided for verification.

In this paper, we propose a wavelet-transform-based watermarking method by adding visually recognizable image to the large coefficients at all frequency bands of the DWT of an image. The first advantage is that the extracted watermark is visually recognizable to claim one's ownership. The second advantage is that the watermarking method is hierarchical and has multiresolution characteristics. The third advantage is that the human eyes are not sensitive to the small changes in edges of an

image. The large coefficients in these bands usually indicate edges in an image. Therefore, it is difficult for human to perceive when adding watermarks on these large coefficients. The fourth advantage is that this approach matches the upcoming image/video compression standards. The watermark is then embedded by modifying significant coefficients at the coarser scale in perceptually important spectral components of image. Our experimental results show that the watermarking method we propose is very robust to image compression and complex image distortions.

2. PRELIMINARIES

The wavelet transform is identical to a hierarchical subband system, where the subbands are logarithmically spaced in frequency. The basic idea in the DWT for a two dimensional image is as follows. An image is first decomposed into four parts of high, middle, and low frequencies, i.e., LL1, HL1, LH1, HH1 subbands, by cascading horizontal and vertical two channel critically subsampled filter banks. The subbands labeled HL1, LH1, and HH1 represent the finest scale wavelet coefficients. To obtain the next coarser scale of wavelet coefficients, the subband LL1 is further decomposed and critically subsampled. This process is continued an arbitrary number of times, which is determined by the application at hand. Fig. 1 shows an image is decomposed into ten subbands for three levels. Each level has various band-information such as low-low frequency band, low-high frequency band, high-low frequency band, and high-high frequency band.

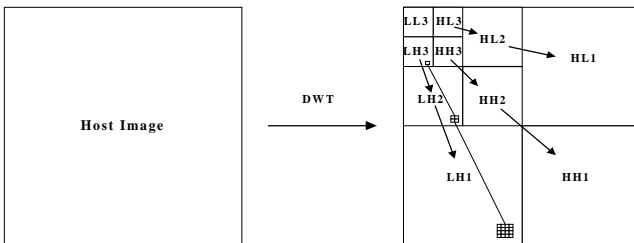
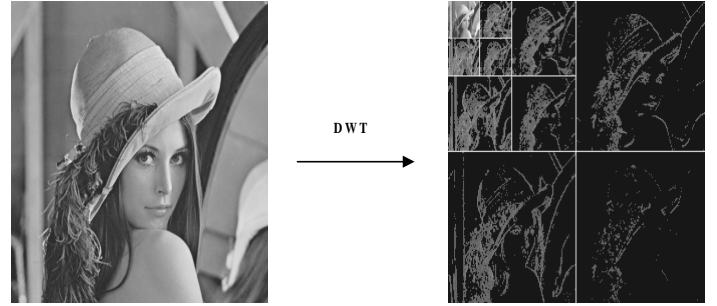


Fig. 1. DWT decomposition of an image: Note that the arrow points from the subband of the parents to the subband of the children. The lowest frequency subband is the top left, and the highest frequency subband is at the bottom right. Also shown is a wavelet tree consisting of all the descendants of a single coefficient in the

subband LH3.

Furthermore, from these DWT coefficients, the original image can be reconstructed. This reconstruction process is called the inverse DWT (IDWT). Let $I[m, n]$ represent an image. The DWT and IDWT for $I[m, n]$ can be similarly defined by implementing the DWT and IDWT for each dimension m and n separately: $DWT_n[DWT_m I[m, n]]$. Fig. 2 shows the original and DWT decomposition of 512×512 Lena image.



(a)

Fig. 2. (a) The original and (b) DWT decomposition of 512×512 Lena image.

3. THE WATERMARKING ALGORITHM

The proposed embedded digital watermarking can hide visually recognizable patterns into the image. The goal of digital watermarking is invisible to human eyes but also robust under different attacks. In the proposed method, watermarks are redundantly embedded into the host image by modifying them in the location of context-based wavelet coefficients (CBWC). The magnitude set of M of the current pixel $c_{x,y}$ is estimated using the weighed values of bucket coefficients that have been entropy encoded in the current subband. The energy of a context-based wavelet coefficient is defined as follows.

$$E_{x,y} = \frac{\sum \log(c)}{24} * 0.4 + \log(c_{x,y}) * 0.6 \quad (1)$$

where $E_{x,y}$ is the energy of the center coefficient $c_{x,y}$ of size 5×5 block coefficients.

3.1 Watermark Embedding Method

The algorithm to embed watermark into the host image is as follows:

Step1: Let I be a greyscale image or the luminance component of a color image. In the embedding part, we first decompose the host image to produce a sequence of three

levels with ten subbands of a pyramid structure MI , corresponding to the vertical, horizontal and diagonal details at each of the k resolution levels, and a gross approximation of the image at the coarsest resolution level. We denote the detail image component at the k th resolution level of the host by HL_k , LH_k , HH_k and $k = 1, 2, 3$. The gross approximation is represented by LL_3 .

Step2: Let watermark W be a greyscale image to be embedded. Firstly, an index rank MW is recorded after sorting the coefficients of W in the spatial domain in descending order to produce a sequence of W' such that

$$W'(1,1) \geq W'(1,2) \geq \dots \geq W'(1, n) \geq \dots \geq W'(m, n) \quad (2)$$

Where (i, j) is a coordinate at W' . Then, we compute the third-level discrete wavelet decomposition of W' . Consider each resolution subband s at scale n , sort the detail coefficients in descending order so that $V_s(MWa_{sn}(p, q))$ is a coefficient such that

$$V_{sn}(MWa_{sn}(1, 1)) \geq \dots \geq V_{sn}(MWa_{sn}(1, q)) \geq \dots \geq V_{sn}(MWa_{sn}(p, q)) \quad (3)$$

Where $sn \in \{LL_3, HL_{1,2,3}, LH_{1,2,3}, HH_{1,2,3}\}$. To embed the watermark, consider each resolution subband s at scale n , sort the absolute values of the detail coefficients in descending order so that $V_{sn}(MWb_{sn}(p, q))$ is a coefficient such that

$$V_{sn}(MWb_{sn}(1, 1)) \geq \dots \geq V_{sn}(MWb_{sn}(1, q)) \geq \dots \geq V_{sn}(MWb_{sn}(p, q)) \quad (4)$$

Where $sn \in \{LL_3, HL_{1,2,3}, LH_{1,2,3}, HH_{1,2,3}\}$.

Step3: To embed the watermark, calculate the energy of CBWC of each subband of the third-level discrete wavelet decomposition of host image I to produce a sequence of MI . Consider each image component at the n -th resolution level, choose top N_{sn} of $CBWC_n$, where N_{sn} is the number of watermark coefficients at the n -th resolution component and quantization bounds. A quantization bound is a coefficient to be recognized without casting watermarks. Let the number of watermark coefficients to be embedded into one subband is K_{sn} and a fixed quantization range is nine, then N_{sn}

coefficients are chosen at the component of host image as follows.

$N_{sn} = \text{int}(K_{sn} / 9) + 1$ //fixed quantization blocks

If $K_{sn} \bmod 9 \neq 0$ then

$$N_{sn} = N_{sn} + 9 - (K_{sn} \bmod 9)$$

$$N_{sn} = N_{sn} + 1$$

End

$$N_{sn} = N_{sn} + K_{sn}$$

After that, sort the absolute values of top N_{sn} in descending order and record the index so that $X_{sn}(MI_{sn}(i))$ is a coefficient such that

$$X_{sn}(MI_{sn}(1)) \geq X_{sn}(MI_{sn}(2)) \geq \dots \geq X_{sn}(MI_{sn}(i)) \geq \dots \geq X_{sn}(MI_{sn}(N_{sn})) \quad (5)$$

Where $sn \in \{LL_3, HL_{1,2,3}, LH_{1,2,3}, HH_{1,2,3}\}$.

To extract the watermark without original image, quantize CBWC. In the quantization stage, let $X_{sn}(MI_{sn}(i))$ and $X_{sn}(MI_{sn}(i+10))$ be quantization bounds over coefficients range $X_{sn}(MI_{sn}(i+1))$ and $X_{sn}(MI_{sn}(i+9))$, a quantization value $(|X_{sn}(MI_{sn}(i))| + |X_{sn}(MI_{sn}(i+10))|)/2$ is set to the ranges. An example is illustrated as follows.

Let the number of watermark coefficients to be embedded into a subband is $K_{sn} = 26$, a fixed quantization range is nine, then $N_{sn} = 31$ are chosen at the component of host image. Consider a sequence at a corresponding subband to be cast,

$$X_{sn}(MI_{sn}) = \{1, 2, 3, 5, 7, 9, 11, 13, 15, 16, 17, 24, 28, 29, 30, 31, 32, 33, 36, 39, 41, 42, 47, 48, 50, 51, 52, 55, 56, 58, 61\},$$

After quantization, we have a new sequence,

$$X_{sn}(MI_{sn}) = \{1, 9, 9, 9, 9, 9, 9, 9, 9, 17, 29, 29, 29, 29, 29, 29, 29, 41, 51, 51, 51, 51, 51, 51, 51, 51, 51, 51, 61\}$$

Restore the sequence of key quantization bound $\{1, 17, 41, 61\}$ can easily determine the quantized coefficients value to be cast.

Step 4: The equations used for watermark casting are as follows.

```

index = 1
For i = 1 to Nsn - 10 step 10
  If i = Nsn - 10
    For p = i + 1 to (Ksn mod 9)
      Xsn(MIsn(p)) = Xsn(MIsn(p)) + αs Vsn(MWAsn(p))
    Else
      For p = i + 1 to (i + 9), q = index + 1 to index + 9
        Xsn(MIsn(p)) =
          Xsn(MIsn(p)) + αs Vsn(MWAsn(q))
      End
      index = index + 9
    End
  End

```

Where $X_{sn}(MI_{sn}(p))$ is a coefficient at sn subband, c denotes coefficient of watermark at sn subband, and $\alpha_{s|s=LL,HL,LH,HH}$ indicates scaling parameter to adjust casting energy. In the casting stage, watermarks $V_{sn}(MWA_{sn}(p))$ are redundantly embedded into top $MI_{sn}(N_{sn})$ of corresponding component at the n -th resolution level, $CBWC_{sn}$, where $sn \in \{HL_{1-3}, LH_{1-3}, HH_{1-3}\}$. Watermark coefficients at the gross approximation is randomized before casting to $CBWC_{LL3}$ for robustness.

Step5: Save the scaling parameter $\alpha_{s|s=LL, HL, LH, HH}$, MW , $MWA_{sn|s=LL,HH,HL,LH}$, $n=1,2,3$, $MWB_{sn|s=LL,HH,HL,LH}$, $n=1,2,3$, and $MI_{sn|s=LL,HH,HL,LH}$, $n=1,2,3$ as key indexes to extract watermarks without original image. Take the two dimensional IDWT of the modified DWT coefficients and the unchanged DWT coefficients can easily form watermarked image.

3.2 Watermark Extracting Method

On the other hand, the watermark are detected by using the embedded position indexes and scaling parameter α after the wavelet decomposition of the watermarked image and the original image, as follows:

Step1: We first decompose a watermarked image, X' with DWT into three levels of ten subbands.

Step2: Restore the scaling parameter α , sorting index of watermark's coefficients MWA_{sn} and sorting index of host image's coefficients. The method to extract watermarks is as follows.

```

index = 1
For i = 1 to Nsn - 10 step 10
  value = (|Xsn(MIsn(i))| + |Xsn(MIsn(i+10))|) / 2
  If i = Nsn - 10
    For p = i + 1 to (Ksn mod 9)
      Vsn(MWAsn(p)) = (Xsn(MIsn(p)) - value) / αs
    Else
      For p = i + 1 to (i + 9), q = index + 1 to index + 9
        Vsn(MWAsn(q)) = (Xsn(MIsn(p)) - value) / αs
      End
      index = index + 9
    End
  End

```

Step4: Take the two dimensional IDWT of V_{sn} , where $sn \in \{HL_{1-3}, LH_{1-3}, HH_{1-3}\}$, we have the extracted watermarks W' .

Step5: In our scheme, the extracted watermark is a visually recognizable image. However, the subjective measurement is dependent on factors such as views. Therefore, we measure the similarity of original watermarks W and extracted watermarks by the standard correlation coefficient as

$$correlation = \frac{\sum (x - \bar{x})(y - \bar{y})}{\sqrt{\sum (x - \bar{x})^2} \sqrt{\sum (y - \bar{y})^2}}$$

4. EXPERIMENTAL RESULTS

Fig. 3 shows the original *lena* image, its watermarked copy and the watermark *pepper* in the spatial domain. We see that the watermarked image is not distinguishable from the original image.

The following shows an example of embedding results, where a grey-level image *pepper* is used as the watermark.



Fig. 3: (a) original *lena* image with size of 512×512 (b) *lena* watermarked with our method, PSNR = 40.9342 (c) original *pepper* image as watermarks with size 64×64 (d) extracting results from (a) without any attack using the proposed method, correlation=0.9990.





4.1. On the Robustness Against JPEG Lossy Compression

Table I shows the extracted results from JPEG compressed version of the watermarked images with compression ratio 9.7, 13.4, 18.3, and 26.2. The quality of watermarked images is

still in good situation even under the high compressed ratio. The extracted watermarks and the original watermarks are with high correlation. In the proposed method, the extracted watermark is visually recognizable under the compression ratio, 26.2.

TABLE I

CHANGES OF CORRELATION UNDER JPEG LOSSY COMPRESSION. AS THE COMPRESSION RATIO INCREASES TO 26.2, THE EXTRACTED WATERMARK IS VISUALLY RECOGNIZABLE.

Compression Ratio	JPEG (CR=9.7)	JPEG (CR=13.4)	JPEG (CR=18.3)	JPEG (CR=26.2)
Extracted watermarks				
Correlation	0.9439	0.9198	0.8915	0.8118





4.2 On the Robustness Against Image Procession Operation

Sharpen operations are used to enhance the subjective quality. Table II shows the extracted results of applying enhanced operation to a watermarked image. The extracted results are still highly similar to the original watermark.

Smoothing operations such as median filter are used to decrease spurious effects that may be present in images from a poor transmission channel. Table II shows the extracted results of applying median filter to a watermarked image. The extracted watermark is still visually recognizable.

TABLE II

THE EXTRACTED OF WATERMARKS, CORRELATION AND ERROR RATE VALUES UNDER SHARPEN AND SMOOTHING OPERATION.

Image processing	Median filter	Sharpen	Blur	Resize
Extracted watermarks				
Correlation	0.9050	0.9298	0.9675	0.5754





4.3 On the Robustness Against JPEG Lossy Compression and Mixed Image Processing

Sharpen and blur operations are used to a watermarked image. Table III shows the extracted results are still highly similar to the original watermark.

Table III also shows the extracted results of applying median filter and sharpen image operations to a watermarked image under the JPEG compression ratio, 26.2, the extracted watermark is still visually recognizable.

TABLE III

THE EXTRACTED OF WATERMARKS, CORRELATION AND ERROR RATE VALUES UNDER SHARPEN PLUS SMOOTHING OPERATION.

Image processing	Sharpen+ median filter	JPEG13.4+sharpen+median filter	JPEG18.3+sharpen+median filter	JPEG26.2+sharpen+median filter
Extracted watermarks				
Correlation	0.8638	0.8703	0.8465	0.7751

5. CONCLUSION

This paper has presented a new technique for embedding visually recognizable grey-level watermark, which is seldom discussed in the literature, into the image based on the context-based wavelet coefficient (CBWC). The embedding and extracting methods of the DWT-based approach have been described scrutinizingly in this paper. The experimental results show that the proposed method provides extra robustness against JPEG compression, image procession operation, and even compound attacks.

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