

# Watermarking Experiments Based On Wavelet Transforms

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## A Proposed Watermarking Scheme

*A general watermarking scheme includes*

- (a) a watermark acquisition: Sampling  $N(0, 1)$  with a *private seed*.
- (b) a watermark embedding: insert the watermark into *wavelet coefficients*.
- (c) a watermark extraction and verification

## Criteria for a Watermark to Meet

- **Transparency:** The watermark should be perceptually invisible or its presence should not be confused with the image being protected.
- **Robustness:** The watermark should still be detected after the image has undergone linear or nonlinear operations (attacks) such as median filtering, cropping, scaling, compression, and enhancement.
- **Capacity:** The watermarking strategy must be of allowing multiple watermarks to be embedded into an image with each image still being independently verifiable.



## Watermarking Embedding (2)

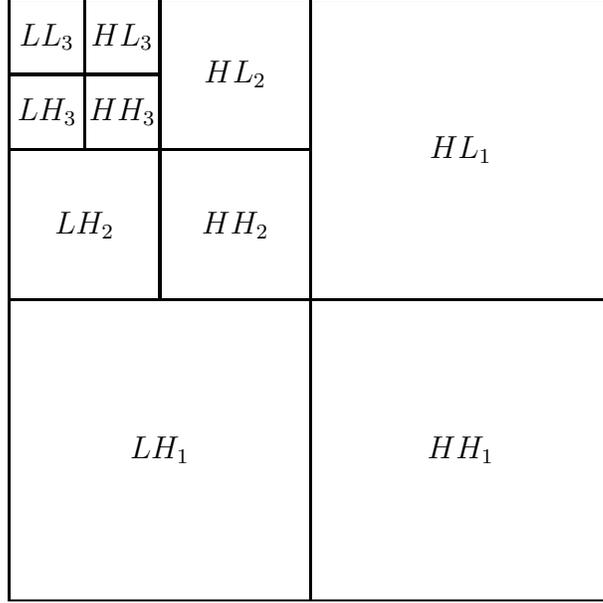


Figure 1: A 3-Scale Wavelet Transform.

where  $Y_1 = HL_3 = H_{HL3}(X)$  and  $Y_2 = LH_3 = H_{LH3}(X)$  are images consisting of wavelet coefficients of high-low and low-high bands at level 3 whose sizes are both  $M_1 \times M_2$ , where  $M_1 = N_1/8$  and  $M_2 = N_2/8$ , respectively. Let  $W$  be a watermark of size  $M_1 \times M_2$  which is acquired by sampling  $N(0, 1)$ , a Gaussian distribution with zero mean and unit variance. Our embedding scheme is done pointwise by

$$Y_1 \leftarrow Y_1 * (1 + \alpha W) \tag{2}$$

$$Y_2 \leftarrow Y_2 * (1 - \alpha W) \tag{3}$$

where  $\alpha \in (0, 0.3]$ .

# Watermarking Extraction and Detection

Let  $\{X(i, j)\}$  be the original image of  $N_1 \times N_2$ , and let  $\{W(i, j)\}$  be an authorized watermark, a matrix of  $M_1 \times M_2$ . Suppose that  $\{Y(i, j)\}$  is an observed image of  $N_1 \times N_2$ , then the extracted watermark  $W^*$  can be computed by the following formulas:

$$Z = H_{HL3}(X) \text{ or } Z' = H_{LH3}(X) \quad (4)$$

$$T = H_{HL3}(Y) \text{ or } T' = H_{LH3}(Y) \quad (5)$$

$$W^*(i, j) = \frac{1}{\alpha}[T(i, j)/Z(i, j) - 1] \text{ or } W^*(i, j) = \frac{-1}{\alpha}[T'(i, j)/Z'(i, j) - 1] \text{ or} \quad (6)$$

$$Sim(W^*, W) = (W^*, W)/\sqrt{(W^*, W^*)} \quad (7)$$

According to a theorem of Probability Theory,  $\{W(i, j)\}$  can be treated as a random sample of size  $K=M_1 \times M_2$  from  $N(0, 1)$ , and  $\{W^*(i, j)\}$  is a set of  $K$  numbers, thus,  $Sim(W^*, W) \sim N(0, 1)$ . Therefore, the two-sided confidence interval of  $Sim(W^*, W)$  is  $[-1.96, 1.96]$ , which helps determine the *significance* of the  $Sim(W^*, W)$  index or the existence of an extracted watermark.

## Experimental Results

Haar	Exp-1	Exp-2	Exp-3	Exp-4	Exp-5	Exp-6	Exp-7	Exp-8
PSNR	41.29	30.46	33.96	45.95	33.34	32.60	30.51	36.01
Sim	41.95	2.66	3.33	23.17	3.40	3.41	2.91	3.12

Table 1: PSNR and Sim of Watermarked Lenna, lena0, Under Attacks.



(a)



(b)

Figure 2: (a) Lenna, (b) lena0: A Watermarked Image.

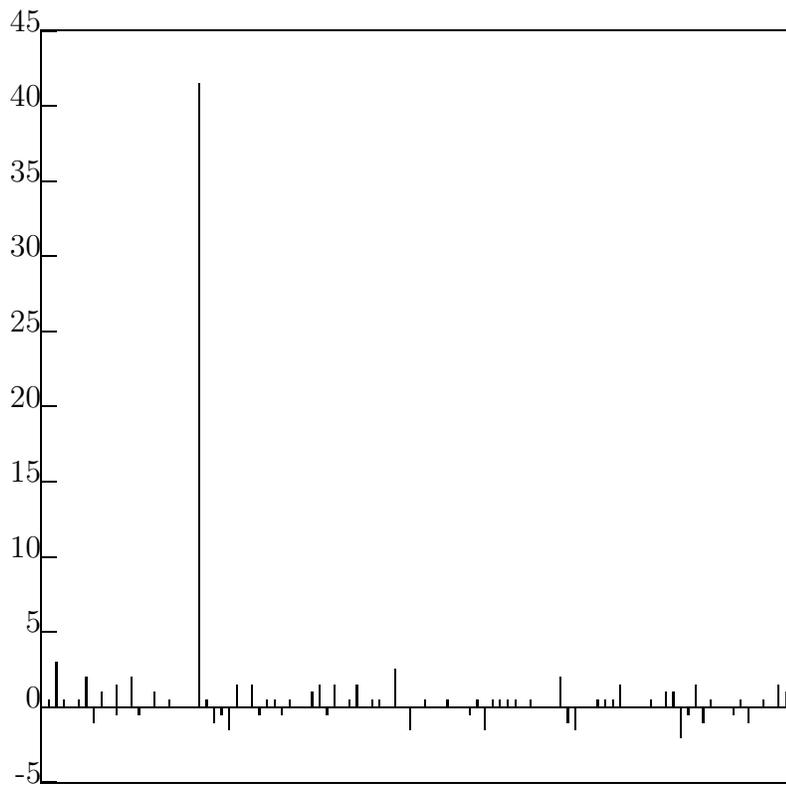


Figure 3: 99 Sim values of random watermarks vs. a true one.



Figure 4: Scaling



Figure 5: Smoothing



Figure 6: Cropping



Figure 7: Noise-Adding



Figure 8: JPEG Compression



Figure 9: Fractal Compression



Figure 9: Wavelet Compression