# Wavelet Transform based Estimation of Images using different Thresholding Techniques

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# ABSTRACT

Estimating the images using decimated wavelet transform is very popular technique in different applications. In this paper a new thresholding function with combination of Smoothly Clipped Absolute Deviation (SCAD), Hard thresholding and soft thresholding functions are introduced for wavelet based denoising of images. The proposed technique is applied for denoising of noisy images contaminated with additive white Gaussian noise using Top rule and Visu rule. The results are compared with that of existing SCAD, hard and soft functions denoising method. Root Mean Square Error (RMSE) and Peak Signal to Noise Ratio (PSNR) are used as parameters for testing the quality of denoising.

#### **Keywords**

Wavelet transform, decimated wavelet transform, image denoising, new thresholding function, top rule, Visu rule.

# 1. INTRODUCTION

Images are acquired in many applications like satellite communication, medical diagnosis. They will be transmitted from one place to another place. In this process noise will be added to the image. This noise affects the clarity of the image at the receiver. To get the original image, denoising of the image is required. Wavelet transform, due to its excellent localization property, has rapidly become an essential signal and image processing tool for a variety of applications, including denoising and compression. Wavelet denoising attempts to remove the noise present in the signal while maintaining the signal characteristics, regardless of its frequency content. Different methods are used for image denoising. The popular techniques for denoising of images are based on wavelet transform. Out of these methods, denoising of image using thresholding method is widely used technique. This method is proposed in this project. In this method

wavelet transform is applied on the noisy image. The outputs of wavelet transform are the noisy wavelet coefficients. Then the threshold value is fixed by applying thresholding rule. The wavelet coefficients are changed using the threshold function. The output of the function is applied to the inverse wavelet transform. It gives the de-noised image. This method is implemented with different noise levels, different images, and different wavelets. These results are compared across the different methods.

# 2. IMAGE DENOSING

In the field of image processing major problem occurred with noisy image. Image denoising is to recover an original image from an observed noisy image. The wavelet thresholding for image de-noising involves the following steps

1. To get the wavelet coefficients by using the wavelet decomposition on the input noisy image.

#### p=P(B+N)

Where p is the wavelet coefficients, P is the wavelet transform, N is the noise data, and B is the original image.

- 2. Apply the thresholding function by using the thresholding rules, on the wavelet co- efficient.
  - **Ρ'=**δ(p)

Where P' is the optimal estimation of the wavelet coefficients,  $\delta(p)$  wavelet thresholding function ,  $\Box$  is threshold.

3. Then apply the inverse wavelet transform on the modified wavelet coefficients and get the de-noised image

B'=P-1p'



Fig 4.1: Block Diagram

# **3. THRESHOLDING RULES**

#### 3.1 Top Rule

Top rule is a global thresholding method and it is independent of thresholding function selected. Given Z as the fraction of the largest coefficients to keep, the threshold is set to be the (1-Z) th quintile of empirical distribution of absolute values of wavelet coefficients.

#### 3.2 Visu Rule

This rule was developed by donoho and Johnston. This is a universal thresholding method not depending on thresholding filter selected. The threshold value is calculated by

$$\lambda = \sigma \sqrt{2 \log L}$$

where L is length of the image,  $\sigma$  is the standard deviation, and  $\lambda$  is the threshold value.

#### 4. THRESHOLDING FUNCTIONS

Thresholding is a simple non-linear technique, which operates on one wavelet coefficient at a time. In its most basic form, each coefficient is threshold by comparing against threshold. If the coefficient is smaller than threshold then it is set to be zero otherwise it is kept or modified. Then inverse wavelet transform will be applied on the result which leads to reconstruction with the essential signal characteristics and with less noise. Since the work of Donoho and Johnstone, there has been much research on finding thresholds, however few are specifically designed for images.

# 4.1 Soft thresholding function

If the coefficient value (p) is greater than the threshold value  $(\lambda)$ , then it taken as the difference between coefficient value and threshold value otherwise the value is zero.

$$S(p,\lambda) = \begin{cases} sgn(p)(|p| - |\lambda|) & for \ |p| > \lambda \\ 0 & otherwise \end{cases}$$

#### 4.2 SCAD function

$$SCAD(p, \lambda) = \begin{cases} sign(p)max(0, |p| - \lambda) \\ if |p| \le 2\lambda \\ (\alpha - 1)p - \alpha\lambda sign(p) \\ if 2\lambda \le |p| < \alpha\lambda \\ p & if |p| > \alpha\lambda \end{cases}$$

Where  $\alpha$ =3.7, SCAD means smoothly clipped absolute deviation.

#### 4.3 Hard thresholding function

Donoho and Johnston proposed hard thresholding filter. A threshold value is selected by adopting some sort of threshold rules. Hard thresholding sets any coefficient below or equal to the threshold to zero. The coefficients above this threshold value are retained. The hard-thresholding "H (e,  $\lambda$ )" is denoted by

H (e, 
$$\lambda$$
) = e for  $|e| > \lambda$   
= 0 otherwise

#### 4.4 New thresholding function

It is developed based on the median of the SCAD function, hard function and soft thresholding function. If the coefficient value (p) is greater than the thresholding value ( $\lambda$ ), then it allows the function value, otherwise it's taken the twenty percentage of coefficient value with output of function.

# 5. EXPERIMENTAL RESULTS AND DISCUSSION

The performance of the proposed thresholding function is evaluated and compared with soft hard and SCAD thresholding functions using wavelets. The experiment is conducted on the image 512 X 512 of size. The Gaussian noise with different variations added to the input image. By applying the wavelets transforms to get the wavelet coefficients those coefficients are modified according in to the shrinkage function and apply the inverse wavelet transforms on the image finally get the original image or denoising image. The performance of the denoising is evaluated by using the PSNR & RMSE.

RMSE is defined as root mean squared error between original image and denoising image

RMSE = 
$$\sqrt{\frac{1}{n} \sum_{i=1}^{n} (X(i) - \hat{X}(i))^2}$$

PSNR=20 log10 (255/square root of MSE)

Where X(i) the original image is  $\hat{X}(i)$  is the de-noised image, n is number of samples. The simulation experiment is repeated 100 times and average values are taken the process is conducted on different images and the results are same. This simulation is developed in MATLAB environment. The results of images for  $\sigma$ =10, 20 and 30 using Soft,hard, SCAD and new thresholding filter with visu rule method and top rule method are shown in Table 1 and Table2. The original and denoised images using new thresholding filter with visu rule method and Top rule method are shown in Figs 1-4. Graph 1-4 shows the comparison of the results of visu rule and top rule methods.

For  $\sigma$ =10, RMSE is 10.8689 and PSNR is 27.4071 are obtained on denoising of noisy image with SCAD filter and RMSE is 12.8571 and PSNR is 25.9480 are obtained with soft thresholding function and RMSE is 8.0409 and PSNR is 30.0247 are obtained with hard thresholding function using visu rule method (Table1). For new thresholding function RMSE of 8.8139 and PSNR of 29.2275 are obtained (Table 1). From the above values we can observe the new thresholding function performs better than the SCAD and soft thresholding function. Similarly performance is obtained for the remaining values  $\sigma$ =20 and 30 (Table 1).

Using Top Rule method, for  $\sigma =10$ , RMSE of 8.6454 and PSNR of 32.3137 are obtained on denoising of noisy image with SCAD thresholding filter and RMSE of 7.9451 and PSNR of 30.1288 are obtained with soft thresholding filter and RMSE of 8.8571 and PSNR of 29.1850 are obtained with hard thresholding filter. For new thresholding filter, RMSE of 7.2633 and PSNR of 30.9081 are obtained (Table 2). This indicates that the new thresholding filter performs better than both SCAD, hard and soft thresholding filters. For  $\sigma =20$  and 30 the new thresholding function gives better results than the SCAD, hard functions only.

	σ=10		σ=20		σ=30	
	RMSE	PSNR	RMSE	PSNR	RMSE	PSNR
Noisy Image	9.9616	28.1642	19.9724	22.1222	29.9724	18.5964
Estimation using Soft function	12.8571	25.9480	17.6203	23.2106	18.3817	22.8431
Estimation using SCAD function	10.8689	27.4071	16.4470	23.8091	18.4187	22.8256
Estimation using Hard function	8.0409	30.0247	11.3460	27.0339	15.8616	24.1238
Estimation using New shrinkage function	8.8139	29.2275	13.3936	25.5928	16.0211	24.0369

#### Table 1: Denoising Results of Fingerprint image using SCAD, hard, Soft and new Thresholding Function: Visu rule.

#### Table 2: Denoising Results of Fingerprint image using SCAD, hard, Soft and new Thresholding Function: Top Rule

	σ=10		σ=20		σ=30	
	RMSE	PSNR	RMSE	PSNR	RMSE	PSNR
Noisy Image	9.9616	28.1642	19.9724	22.1222	29.9724	18.5964
Estimation using Soft function	7.9451	30.1288	11.2000	27.1464	14.3554	24.9905
Estimation using SCAD function	8.6454	32.3137	13.0913	25.7911	17.7684	23.1378
Estimation using Hard function	8.8571	29.1850	17.3234	23.3581	25.4437	20.0192
Estimation using New shrinkage function	7.2633	30.9081	11.4354	26.9658	15.6758	24.2262



Fig 2: Nosiy image (σ=20)



Fig 3: denoised image using new thresholding function with visu rule



Fig 4: denodied image using new thresholding function with Top rule



Graph 1: RMSE values of different functions in Visu rule method



Graph 3: RMSE values of different functions in Top rule method

# 6. CONCLUSION

In this paper, a new thresholding filter for wavelet shrinkage denoising of image is proposed. The performance of this filter is evaluated by using Fingerprint image. The results are compared with existing SCAD, hard and soft filters. It is found that the new thresholding filter performs better than both SCAD and Soft filters with visu rule method and performs better than SCAD and hard function with Top Rule. In this project a new thresholding filter is used for the denoising of black and white images in future it may be able to apply for the colour images and videos also.

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Graph 2: PSNR values of different functions in visu rule method



Graph 4: PSNR values of different functions in Top rule method

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