# An Enhanced Audio Noise Removal Based on Wavelet Transform and Filters.

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## **Abstract**

The recovery of a kind signals from the correspondence channel with no noise is a troublesome task. Numerous procedures has develop by the interest of expulsion about the noises for a digital signal. One of affective procedure is wavelet denoising, this procedure is depending on threshold algorithm which is an intense strategy for stifling noise in digital signals. In this, the system depends on DDDTWT utilizing a level banks subordinate threshold algorithm is actualized in it. The input signal is included along Additive White Gaussian Noise which is to be decided during execution. The output is regarding SNR and RMSE which are contrasted and the estimations of DTDWT and DDDWT methods and furthermore with a threshold value. The results are appeared in MATLAB simulations demonstrate that the proposed technique is more successful and this gives better execution for denoising the audio signals as far as the SNR and RNISE.

**Keywords:** Signal to Noise Ratio; Root Mean Square Error; Double Density Dual Tree Discrete Wavelet Transform; Dual-Tree Discrete Wavelet Transform; Double-Density Discrete Wavelet Transform.

## I. INTRODUCTION

Noise diminishment in discourse signals is a field of concentrate to recoup a unique flag from its noise debased flag. The noise can be of different sorts like white, impulsive or even different sorts of noise generally found in discourse signals. Over the previous decades, the expulsion of this noise from audio signals has turned into a territory of enthusiasm of a few scientists around the globe, since the nearness of noise can

altogether debase the quality and intelligibility of these signals. Many examinations have been directed since the sixties, with the objective of creating calculations for enhancing the quality of audio and discourse. Ching-Ta and Hsiao-Chuan Wang have proposed a method in view of basic band decay which converts a noisy flag into wavelet coefficients (WCs), and upgraded the WCs by subtracting a limit from noisy WCs in each sub band.

Time recurrence based examination of discourse flag has been presented by Marián Képesia and Luis Weruaga. In this Short-Time Fan-Chirp Transform (FChT) was characterized univocally. Nanshan Li et al. have proposed an audio denoising calculation on the premise of versatile wavelet delicate edge which depends on the pickup factor of straight channel framework in the wavelet space and the wavelet coefficients trigger vitality administrator with a specific end goal to advance the impact of the substance based tunes recovery framework. Eric Martin have presents a versatile audio piece thresholding calculation. The denoising parameters are processed by the time-recurrence normality of the audio flag utilizing the SURE (Stein Unbiased Risk Estimate) hypothesis. B. JaiShankar and K. Duraiswamy have presented the noises introduce in correspondence channels are exasperating and the recuperation of the first signals from the way with no noise is exceptionally troublesome undertaking.

It gives a good representation compared to Gaussian distribution function with regard to filtering of noise from the signal. The signal denoising methods based on D<sup>2</sup>DWT and DTDWT where implemented in results. The optimal values of starting point and decomposition level were determined. These denoising methods give more achievement in charge of SNR and RMSE when compared with the conventional DWT method. In this, audio denoising technique depends D<sup>2</sup>DTDWT using a level dependent threshold is implemented. The threshold points and the decomposition levels for the wavelet denoising methods based on the strength of the noise involved in the signal. In the proposed work, threshold in VisuShrink method or universal threshold introduced by Donoho is modified and applied to various stages of D<sup>2</sup>DTDWT.

## II. METHODOLOGY USED

## A. Wave File Format

Waveform Audio File Format (WAVE, or all the more generally known as WAV because of its filename extension), (also, yet once in a while, named, Audio for Windows) is a Microsoft and IBM audio file format standard for putting away an audio bit stream on PCs. It is an application of the Resource Interchange File Format (RIFF) bit stream format technique for putting away information in "lumps", and in this manner is additionally near the 8SVX and the AIFF format utilized on Amiga and Macintosh PCs, respectively. It is the principle format utilized on Windows frameworks for crude and ordinarily uncompressed audio. The standard thing bit stream encoding is the linear

pulse-code modulation (LPCM) format.

## B. Wavelet Based De-noising System

There has been a decent lot of research on wavelet-based picture denoising. The paper distributed by Donoho and Johnston (1994), built up a hypothetical system for denoising signals utilizing Discrete Wavelet Transform (DWT). The technique comprises of applying the DWT to the first information, thresholding the definite wavelet coefficients and converse transforming the arrangement of edge coefficients to acquire the de-noised signal.

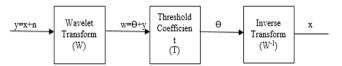


Figure 1: Block Diagram of DWT

Given a loud flag y = x + n where x is the coveted flag and n is free and indistinguishably circulated (i.e.) Gaussian clamor N (0,  $\sigma$ 2), y is first decayed into an arrangement of wavelet coefficients w = W[y] consisting of the coveted coefficient q and commotion coefficient n. By applying a reasonable limit esteem T to the wavelet coefficients, the coveted coefficient q = T[w] can be acquired; in conclusion a backwards change on the coveted coefficient q will produce the denoised flag  $x=w-1[\theta]$  [10].

To realize the twofold thickness twofold tree DWT, we should first framework a legitimate channel bank structure (one that solidifies the characteristics of the twofold thickness and twofold tree DWTs). We have seen what kind of channel bank structure is connected with the twofold thickness DWT in the past zones (generally that it is made out of one low pass scaling channel and two high pass wavelet channels), so we will now swing to the properties of the twofold tree DWT. Therefore, the channel bank structure contrasting with the twofold thickness complex DWT contains two oversampled iterated channel banks working in parallel on a comparable data. The iterated oversampled channel bank coordinate, identifying with the simultaneous execution of the twofold thickness and twofold tree DWTs, is appeared in Figure 2 underneath.

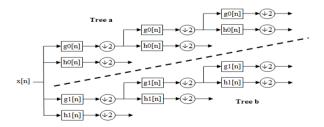


Figure 2: Iterated Filter bank for the Double-Density Complex DWT

#### III. ADVANCED METHOD

Threshold algorithm for wavelet transform is the very powerful technique to compress noise in digital signals. An audio signal which contains Gaussian noise is used for the procedure. The compress levels and threshold value of the denoising methods based on the intensity of the noise implemented in the signal.

The procedure involves in the audio denoising process are described below.

- 1) Read an audio signal s (i).
- 2) Add Gaussian noise to this signal to form noised signal r (i). The best technique is to test the effect of noise on a signal is to add Additive White Gaussian Noise.

The noised signal is in the form of equation (1).

$$r(i) = s(i) + \sigma \varepsilon(i), i = 0,1,2...,n \tag{1}$$

Where r (i) is the noisy signal, s (i) is the noise-free signal to be detected, (i) is the noise signal, o - noise intensity and n is the length of the signal.

3) Tally the signal to noise ratio in decibels and root means square error of the noisy signal using the formulae given by equations (2) and (3) respectively.

$$SNR(dB) = 10 \log_{10} \left[ \frac{\sum_{i=1}^{n} x_i^2}{\sum_{i=1}^{n} (r_i - s_i)^2} \right]$$
 (2)

$$RMSE = \sqrt{\frac{\sum_{i=1}^{n} (r_i - s_i)^2}{n}}$$
 (3)

Where s is the starting signal, r is the noisy signal/ reconstructed signal and n is the magnitude of the signal.

4) For the observed SNR/RMSE of the received signal, compute the threshold value for wavelet thresholding algorithm. The level dependent threshold value is computed using the modified form of the universal threshold introduced by Donoho as given by equation (4).

$$T_{new} = \sqrt[\sigma]{2\log\frac{N}{2^I}} \tag{4}$$

Where  $\sigma$  is the noise intensity, N represents the number of samples and I is the decomposition stage. The global threshold algorithm uses the universal threshold given by equation (5).

$$T = \sqrt[\sigma]{2 \log N} \tag{5}$$

5) To decompose the received signal into wavelet coefficients using the corresponding analysis filter banks for DDDTDWT/DDDWT/DTDWT staffing from the first stage. Utilize different filter bank for all the levels.

6) Threshold wavelet coefficients: Process each sub band separately in a loop. Apply threshold value to the decomposed wavelet coefficients through all scales and sub bands using the level dependent and global threshold values. Selecting the threshold value is more important in wavelet threshold denoising method. The selection of threshold value is too small or too large, the signal cannot be accurately measured. Soft and hard threshold are the two types of thresholding signal used in wavelet denoising and they are given by equations (6) and (7).

Soft threshold: 
$$\{r = sign(s)(|s| - T)\}$$
 (6)

$$Hard\ threshold: \begin{cases} r = s, if \ |s| > T \\ r = 0, if \ |s| < T \end{cases} \tag{7}$$

Where s will be the input signal, r will be the signal after applying threshold signal and T is the threshold point.

- 7) Reconstruct the signal: Consider the reconstruction using the threshold wavelet coefficients with the synthesis filter banks for DDDTDWT DDDWT DTDWT.
- 8) Calculate SNR and RMSE of the denoised signal using equations (2) and (3). The obtained values are then compared with the values of previous step 2 for grading the performance. Higher the SNR or lower the RMSE will decide the realization of the denoising method.
- 9) Repeat the above steps for different decomposition stages. For each stage, the SNWRMSE is calculated. The decomposition stage that gives highest SNR or least RMSE is the optimum stage. Figure 3 shows the necessary steps involved in the denoising process.

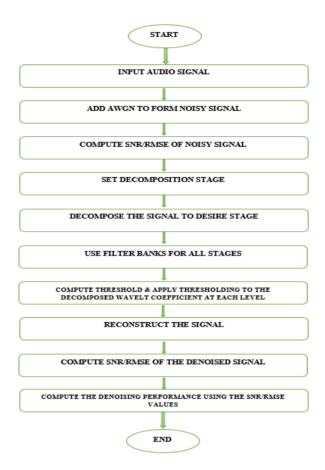


Figure 3: Work flow of the denoising method

## IV. RESULTS

Various threshold rules to be followed are discussed below.

- **1.** '**rigrsure**', adaptive threshold selection using principle of Stein's Unbiased Risk Estimate.
- **2. 'heursure**', heuristic variant of the first option.3
- **3.** 'sqtwolog', threshold is sqrt (2\*log (length(X))).
- **4.** 'minimaxi', minimax thresholding.

The running programming during which threshold rule is selected from four types as: **heursure**, **rigrsure**, **minimax**, **sqtwolog** for all wavelet types filter as well as this selection rule is applied on FIR and median filter Table 1-3, shows the soft thresholding for heursure, rigrsure, minimaxi and sqtwolog respectively.

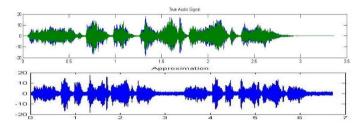


Figure 4: Approximation and True Signal

**Table 1:** Wavelet Transform Techniques Filters (heursure)

S.No:	<b>Techniques Names</b>	Noisy SNR	Denoised SNR	Efficiency
1	db13	5.0046	5.2745	8.885
2	db40	4.8975	5.2822	8.885
3	sym13	5.0059	5.3480	8.885
4	Sym21	5.0001	5.2487	8.885
5	Haar	4.3261	5.3102	8.885
6	Median Filter	5.0012	-6.9215	1.436
7	Adaptive Filter	4.3399	-10.4756	1.424

 Table 2: Wavelet Transform Techniques Filters (rigrsure)

S.No:	<b>Techniques Names</b>	Noisy SNR	Denoised SNR	Efficiency
1	db13	4.0036	6.2745	8.885
2	db40	4.4575	6.2822	8.885
3	sym13	5.0019	6.3480	8.885
4	Sym21	5.0098	6.2487	8.885
5	Haar	5.0261	6.3102	8.885
6	Median Filter	4.9912	-7.9213	1.4369
7	Adaptive Filter	5.0099	-10.4125	1.4245

 Table 3: Wavelet Transform Techniques Filters (minimaxi)

S.No:	<b>Techniques Names</b>	Noisy SNR	Denoised SNR	Efficiency
1	db13	5.0006	5.7745	8.885
2	db40	4.9975	5.7822	8.885
3	sym13	5.0059	5.6480	8.885
4	Sym21	5.0017	5.6487	8.885
5	Haar	4.3261	5.3102	8.885
6	Median Filter	5.0002	-1.9215	2.336
7	Adaptive Filter	5.0099	-4.4756	2.324

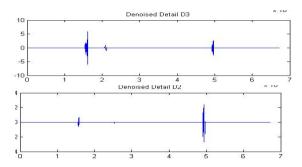


Figure 6: Noise and Denoised Signal at Filter heursure

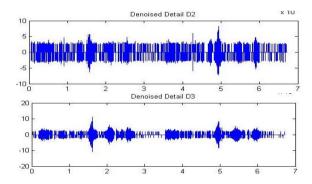


Figure 7: Noise and Denoised Signal at Filter rigrsure

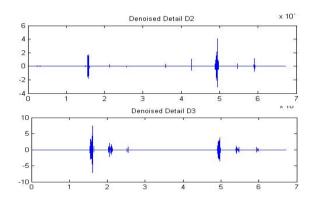


Figure 8: Noise and Denoised Signal at Filter minimaxi

At last the final Noisy and Denoised Audio Signal are as follow, according to below-mentioned value.

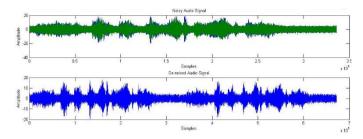


Figure 7: Final Output of Audio Signal

## v. CONCLUSION

Audio denoising is a procedure on base of piece composed strategy. This system depends on the denoising methodology and its effective work was displayed in full detail. The usage comes about has found that the procedure of square planned has accomplished a condition of-heathenizing execution regarding both pinnacle signal-to-clamor proportion and subjective progress in the capable of being heard nature of the audio signal. Gathering of similar pieces enhanced the productive operation of the procedure. The squares were separated and revived in their unique positions from where they were gotten. The gathered squares were lap-jointed each other and along these lines for each and every component, a much variety estimation was found for consolidated to expulsion of clamor from the info signal. The decline in the commotion level translates that the strategy has been secure the essential one of a kind normal for every individual piece despite the fact that the finest points of interest were contributed by gathered squares. What's more, the method can be restricted for different other audio signals and also for different issues that can be valuable for exceedingly straight signal portrayals.

## VI. FUTURE SCOPE

The examination for proficient signal denoising methods is still remain for a substantial test. Wavelet fill in as power all mechanical assembly for the task of signal denoising and the denoising viability can be created by using the expansive sorts of DWT. In this paper, an audio denoising technique in perspective of twofold thickness double tree DWT using a level depended edge calculation is realized. The denoising ability of twofold thickness double tree DWT is then contrasted and the consequences of twofold thickness DWT and double tree DWT strategy. The graphical investigation and forbidden outcomes show that DDDTDWT gives more SNR and less R2v1SE esteems than interchange methodologies. This is proficient at different deterioration stages and clamor levels. The consequences of level ward limit calculation are likewise contrasted and that of worldwide thresholding strategy plainly the denoising strategies and their parameters ought to be picked by the signals close by.

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