Novel Method for Denoising Medical Image Using 2nd Level Discrete Wavelet Transform and Bilinear Filter

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Abstract: Medical imaging is one of the crucial subfields in the world of science and technology. There must be genuine image quality of medical images as it is used to diagnose diverse types of illness and in medical image we cannot settlement or compromise in quality because it provide us important data of patients and if quality compromised then severe effects may come into existence which definitely harmful for patients. Developing a significant denoising method plays a crucial role in image processing. In this research paper, image is first decomposed using filter into eight subbands using 3D DWT and bilateral filter technique and in this process we got approximation coefficient using DWT and once again DWT technique used on approximation coefficient image then we applied bilateral filter and the detail coefficients are subjected to Wavelet Thresholding. After that there is requirement of image reconstructed and this process is executed by inverse wavelet transform (IDWT) of the resultant coefficients and then it is filtered using bilateral filter. In our research work two types of images are considered which are MRI images and Ultrasound images. In this work IDWT process carried out two times that is why our research work known as 2 level DWT process. In this process finally two parameters are calculated PSNR and MSE.

Keywords- PSNR, Wavelet, MSE, MRI, Ultrasound, IEF, Medical, Quantization, Pixel, 3D

I. INTRODUCTION

Digital image processing subcategory of digital signal processing (DSP), digital image processing has lot of attractive features as compared to analog image processing. It allows very wider scope of algorithms to be applied to the input data and can avoid difficulties such as noise in signal and signal variation during processing. Since images are defined over two dimensions. DIP may be design in the form of Multi dimensional Systems. Digital image processing allows the use of complicated algorithms for image processing, and hence, can offer both more sophisticated performance at simple tasks, and the implementation of different approach which would be impossible by analog means. The block diagram of a digital camera is shown in Figure 1. A lens focuses the light from regions of interest onto a sensor. The sensor measures the color and light intensity. An analog-to-digital converter (ADC) converts the image to the digital signal. An image-processing block enhances the image and compensates for some of the deficiencies of the other camera blocks. Memory is present to store the image, while a display may be used to preview it. Some blocks exist for the purpose of user control. Noise is added to the image in the lens, sensor, and ADC as well as in the image processing block itself. The sensor is made of millions of tiny light-sensitive components. They differ in their physical, electrical, and optical properties, which add a signal-independent noise (termed as dark current shot noise), to the acquired image. Another component of shot noise is the photon shot noise. This occurs because the number of photons detected varies across different parts of the sensor. Amplification of sensor signals adds amplification noise, which is Gaussian in nature. The ADC adds thermal as well as quantization noise in the digitization process. The imageprocessing block amplifies part of the noise and adds its own rounding noise. Rounding noise occurs because there are only a finite number of bits to represent the intermediate floating point results during computations [2]. Most denoising algorithms assume zero mean additive white Gaussian noise (AWGN) because it is symmetric, continuous, and has a smooth density distribution.



Figure1 Basic Blocks of a Digital Camera and Possible Sources of Noise

Multi wavelets also improve denoising performance as compared to the scalar wavelet [11]. Wavelet transforms which

generate more wavelet coefficients than the size of the input data are termed redundant or over complete. This added redundancy improves the denoising performance.



Figure2 Multi resolution Analysis (MRA)

II. LITERATURE SURVEY

Martin A. Styner; Elsa D. Angelini [2017]: Automatically detecting anatomy orientation is an important task in medica1 image analysis. Specifically, the ability to automatically detect coarse orientation of structures is useful to minimize the effort of fine/accurate orientation detection algorithms, to initialize non-rigid deformable registration algorithms or to align models to target structures in model-based segmentation algorithms. In this work, we present a deep convolution neural network (DCNN)-based method for fast and robust detection of the coarse structure orientation, i.e., the hemi-sphere where the principal axis of a structure lies. That is, our algorithm predicts whether the principal orientation of a structure is in the northern hemisphere or southern hemisphere, which we will refer to as UP and DOWN, respectively, in the remainder of this manuscript. The only assumption of our method is that the entire structure is located within the scan's field-of-view (FOV). We tested our method on 114 Elbow MR Scans. Experimental results suggest that only five 2-channel images are sufficient to achieve a high success rate of 97.39%. Our method is also extremely fast and takes approximately 50 milliseconds per 3D MR scan. Our method is insensitive to the location of the structure in the FOV [1].

Taruna Grover [2016]: The Medical Images are facing a problem of high level components of noises. There are different techniques for producing medical images such as Magnetic Resonance Imaging (MRI), X-ray, Computed Tomography and Ultrasound, when this process occurs noise is added that decreases the image quality and image analysis. Denoising techniques are used to remove the noise or distortion from images while preserving the original quality of the image while wavelet transform improves the quality of an image and reduce the noise 1eve1. The aim of this paper to characterize the Gaussian noise in wavelet transforms. Subsequently a threshold based denoising algorithm has been developed using hard and soft thresholding techniques. It works on Haar, Daubechies and Sym1et Transforms. First1y the image is decomposed using Haar and Daubechies and sym1et transforms, and then the 1eve1 of soft and hard threshold is selected for reducing the noise in the image and after that by calculating and comparing the PSNR& MSE of an image for every wavelet. Haar transform decomposed the discrete signal into two sub signals of half its length. One sub signal is a running average or trend and other is running difference or fluctuation. Daubechies wavelet has set of scaling functions that are orthogonal. It is useful in noise removal as high frequency coefficient spectrum reflect all high frequency changes. Hard thresholding is a keep or kill procedure. Soft thresh holding shrinks the coefficients above the threshold in absolute value [2]

Yogesh S. Bahendwar, G.R.Sinha, G.R.Sinha [2015]: A11 digital images contain some degree of noise due to the corruption in its acquisition and transmission by various effects. Particularly, medical image are likely disturbed by a complex type of addition noise depending on the devices which are used to capture or store them. No medical imaging devices are noise free. The most commonly used medical images are received from MRI (Magnetic Resonance Imaging), CT (Computed Tomography) and X-ray equipments. Usually, the addition noise into medical image reduces the visual quality that complicates diagnosis and treatment. Additive random noise can easily be removed using simple threshold methods. This paper proposes a medical image denoising algorithm using Discrete Wavelet Transform (DWT). Numerical results show that the algorithm can obtained higher peak signal to noise ratio (PSNR) through wavelet based denoising algorithm for Medical images corrupted with random noise [3].

Indulekha N R, M Sasikumar [2015]: Medical imaging is one of the most important sub-fields in the world of science and technology. There is no compromise in the quality of medical images as it is used to diagnose a variety of illness. Developing a significant denoising method plays a major role in image processing. In this paper, image is first decomposed into eight subbands using 3D DWT and bilateral filter and Thresholding methods are incorporated. The approximation coefficient obtained from DWT is filtered using Bilteral filter and the detail coefficients are subjected to Wavelet Thresholding. Hard thresholding and Soft threshold are the commonly used thresholding techniques. Image is reconstructed by the inverse wavelet transform of the resultant coefficients and then it is filtered using Bilateral filter. MRI images and Ultrasound images are taken as datasets for quantitative validation. The Peak Signal to Noise Ratio (PSNR), Root mean square error (RMSE), Structural Similarity Index (SSIM) are employed to quantify the performance of denoising [4].

Sumit Kumar Mantosh Biswas [2015]: The term Curvelet transform in the field of Image Processing is quite well known from past few years. Its ability to detect curved features and smooth areas in an image marks its huge importance in the area of image denoising. However the ability to denoise image depends upon the selection and application of threshold after doing Curvelet based decomposition of an image. In this paper we are presenting our research methodology based on Curvelet transform image denoising. Our approach is based on the implementation of a modified window neighborhood processing that adapt itself based on the variance of neighboring pixe1s. We describe the problem we are considering for our research, present a brief overview of relative literature, describe the proposed methodology we have implemented and illustrate our future p1an [5].

Shrestha, Suman [2014]: Noise is a major issue while transferring images through all kinds of electronic communication. One of the most common noises in electronic communication is an impulse noise which is caused by unstable voltage. In this paper, the comparison of known image denoising techniques is discussed and a new technique using the decision based approach has been used for the removal of impulse noise. All these methods can primarily preserve image details while suppressing impulsive noise. The principle of these techniques is at first introduced and then analyzed with various simulation results using MAT1AB. Most of the previous1y known techniques are applicable for the denoising of images corrupted with less noise density. Here a new decision based technique has been presented which shows better performances than those already being used. The comparisons are made based on visual appreciation and further quantitative1y by Mean Square error (MSE) and Peak Signa1 to Noise Ratio (PSNR) of different filtered images [11].

III. METHODOLOGY

In this research work our main goal is to remove various types of noise for example Gaussian, Poisson and Speckle noise from medical image. It's very crucial that noise must be removed efficiently from medical image otherwise consequences will be dangerous for patient if output data wrongly analyzed by doctor. In this research work we need to select a data set of MRI and ULTRASOUND images so that simulation analysis can be done and a comparative analysis can be done with respect to base paper. Our main Objective are:

- Choose appropriate data set
- Apply DWT technique efficiently on reference image so that image can be converted into frequency domain. Because in frequency domain it is very easy to fetch information

Apply filters on frequency domain image:

- Test which filter is best suited among various filter present
- Test for single filter and hybrid filter result

Find out important parameters

- Peak signal to noise ratio
- Root Mean Square Error

The input medical image is corrupted with different types of noise and the noisy input image is transformed into wavelet domain using 3D DWT. After a one-level of 3D discrete wavelet transform, the volume of image data is decomposed into HHH, HHL, HLH, HLL, LHH, LHL, LLH and LLL. The approximate signal, resulting from scaling operations only, goes to the next octave of the 3-D transform.



Figure3 Functional Block Diagram

In our research work our main focus is to compress medical image using DWT technique and analyze the PSNR and MSE value with respect to our base paper and both the parameters must be improved. It has roughly 90% of the total energy. Meanwhile, the other seven sub bands contain the detail signals. The approximate signal band is filtered using bilateral filter which removes the unwanted distortions from it while preserving the fine details and edges. Meanwhile, wavelet thresholding is performed for the remaining 7 detail sub bands. Inverse wavelet transform is applied to the filtered sub bands and image is reconstructed. Finally, the reconstructed image so obtained is passed through bilateral filter obtaining the denoised image. The second objective of the present work is extending the DWT implementation to Diversity Enhanced DWT. Denoising algorithms are implemented with various filtering methods. A new version of hyper analytic wavelet transform (HWT) is implemented with a zero order wiener filtering for image analysis. In the proposed HWT based method the following are the advantages of HWT. With these advantages it is useful in denoising the corrupted images.

Figure4 Denoising using Wavelet Transform Filtering

IV. EXPERIMENTAL RESULT

SOFTWARE: MATLAB R2015A: It is powerful software that provides an environment for numerical computation as well as graphical display of outputs. In Matlab the data input is in the ASCII format as well as binary format. It is high-performance language for technical computing integrates computation, visualization, and programming in a simple way where problems and solutions are expressed in familiar mathematical notation.

WEINER FILTER MSE AND PSNR VALUES:

Mean Square Error = 0.0033 Peak Signal to Noise Ratio = 36.4802



Figure5 MRI Grey Image



Figure6 Low, High, Approximate coefficient after applying first level DWT



Figure7 Low, High, Approximate coefficient after applying second level DWT



Figure8 First level IDWT



Figure9 Second level IDWT

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Figure10 Input image and output of Bilateral Filter



Figur11 Input image corrupted with noise



Figure12 Histogram of denoised Image



Figure13 Output image after applying wiener filter to noised image

V. CONCLUSION

In our research work novel approaches is used for the noise removal or denoise of 3D medical images. In our research work two level DWT technique and bilateral filtering approach is used. In this wavelet thresholding is executed on the sub bands which are obtained from 3D DWT. The technique utilizes the properties of 3D DWT and bilateral filter to contain the information containing pixels & edges and thresholding is performed in order to segment an image by setting all pixels whose intensity values are above a threshold. With pace of time huge development came into existence. Image processing is one of the tremendous filed in which research going on very rapidly. In these days more focus toward medical application which is very helpful in diagnosis of patient. There are various types of images and diverse technology which can be used to process the image in efficient way. In these day machine learning concept is used and besides this artificial intelligence with IoT can boom the application in biomedical field. Hybrid filter can also be used to get better analysis of image.

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