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VLSI Implementation Using Dwt for Image Compression and Image Fusion for Medical Application

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Abstract:

Image fusion is a data fusion technology, which keeps images as main research contents. It is techniques that integrate multi-images of the same scene from multiple image sensor data or integrate multi images of the same scene at different times from one image sensor. The fusion technique in medical images is useful for the resourceful disease diagnosis purpose. For medical diagnosis, Computed Tomography (CT) provides the best information on the denser tissue with less distortion. Magnetic Resonance Image (MRI) provides better information on soft tissue with more distortion. The image fusion algorithm based on Discrete Wavelet Transform which is faster developed was a multi-resolution analysis image fusion method in recent decades. Discrete Wavelet Transform has good time-frequency characteristics. The paper proposes that the method could extract useful information from source images to fused images so that clear images are obtained. In this paper, a hardware implementation of a image fusion system is proposed. Here using MATLAB we can convert images into pixel-format files and to observe simulation results. To implement this paper XPS & VB are needed. Using XPS, first select hardware & software components then by adding source and header files & converting it into bit streams and download into FPGA, to obtain a fused image. The input image can also be recovered by combining of fused image and the other input image.

Keywords: FPGA, MATLAB, Visual Basic etc.

1. Introduction

Image fusion is the process of combining relevant information from two or more images into a single image. Then the resultant image has more information than input images. Image fusion has become a topic of great interest to the engineers working in different fields. It's being used for medical applications, so as to get a better image. It's also being used in automotive industries to enhance the vision of road, to observe a better image during a rainy weather. Due to the increasing number of high-resolution images are available along with sensor technology development, image fusion is still important and easy method to interpret the image data, obtaining a more suitable image for a variety of applications, such as visual interpretation, satellite, , digital classification & war-field. The main objective of medical imaging is to obtain a high resolution image with as much detail as possible for the sake of diagnosis. MRI and CT imaging are of main concern for diagnostic purposes. Both techniques give special sophisticated characteristics of the organ to be imaged. So, it is expected that fusion of MRI and CT images of the same organ would result in an integrated image of much more information. Wavelet transforms fusion is defined as considering the wavelet transforms of the two registered input images together with the fusion rule. Then, the inverse wavelet transform is computed, and the fused image is reconstructed. Generally the image fusion process can be carried out at various levels. Under this, in the pixel-level image fusion the fused images provided all relevant information present in the original images without any inconsistencies[1]. The pixel-level image fusions were classified into spatial domain fusion and transform domain fusion. A Spatial domain fusion is directly applied on the source images which had reduced the signal to-noise ratio of the resultant image with simple averaging technique but the spatial distortion still persists in the fused image. To improve on that in transform domain fusion, firstly the input images are decomposed into some levels based on transforming coefficients. Then the fusion technique is applied and the fusion decision map is obtained. Using Inverse transformation on this decision map gives the fused image. The fused image includes all the details of the source images an and reduces the spatial distortion.

2. Related Work

Over the years, multiple efforts have been made to develop new image fusion techniques. Additionally, efforts have been made to review and assess the image fusion qualities. However, there are still a large number of open issues that need to be resolved in this area. Jasiunas presented an image fusion system based on wavelet decomposition for unmanned airborne vehicles (UAVs). This is probably the first implementation developed on a reconfigurable platform also first investigation adaptive image fusion that makes use of dynamic reconfiguration to change the fusion algorithm as the UAV approaches an object of interest. Sims and Irvine presented an FPGA implementation using pyramid landed composition and subsequent fusion of dual video streams. This realization gives a design that can fuse dual video streams in gray scale video graphics array (VGA), with 30 frames/s in real time. Both hardware designs showed an improvement in speed performance compared to respective algorithms running on a typical PC, giving strong indications of the advantages of hardware against software implementation. Following are the some image fusion technologies having its own advantages and disadvantages A) Fusion using Principle Component Analysis (PCA): The PCA image fusion method simply uses the pixel values of all source images at each pixel location, and adding a weight factor to each pixel value, and takes an average of the weighted pixel values to produce the result for the fused image at the same pixel location. The optimal weighted factors are determined by the PCA techniques [2]. The PCA technique is useful for image encoding, image data compression, image enhancement B) Fusion using Laplacian pyramid method: The Laplacian pyramid fusion consists of an iterative process of calculating the Gaussian and Laplacian pyramids of each source image, fusing the Laplacian images at each pyramid level by selecting the pixel with the larger absolute value, combining the fused Laplacian pyramid with the combined pyramid expanded from the lower level, and then expanding the combined pyramids to the upper level C) Fusion using gradient pyramid method: A gradient pyramid is obtained by applying a set of 4 directional gradient filters (horizontal, vertical, and 2 diagonal) to the Gaussian pyramid at each level. At each level, these 4 directional gradient pyramids are combined together to obtain a combined gradient pyramid that is similar to a Laplacian pyramid. The gradient pyramid fusion is therefore the same as the fusion using the Laplacian pyramid method except replacing the Laplacian pyramid with the combined gradient pyramid.

3. Proposed Model

Image fusion has become a common term used within medical diagnostics and treatment. The term is used when multiple patient images are registered and overlaid or merged to provide additional information. Fused images may be created from multiple images from the same imaging modality, or by combining information from multiple modalities, such as magnetic resonance image (MRI), computed tomography (CT), positron emission tomography (PET), and single photon emission computed tomography (SPECT). The transform of a signal is just another form of representing the signal. It does not change the information content present in the signal. The Wavelet Transform provides a time-frequency representation of the signal. It was developed to overcome the short coming of the Short Time Fourier Transform (STFT), which can also be used to analyze non-stationary signals.

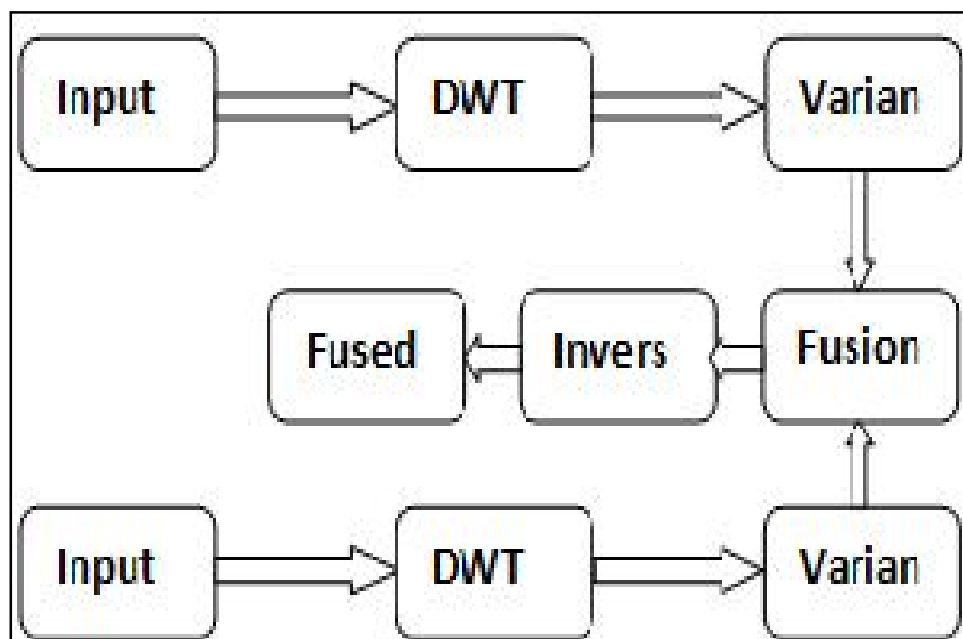


Figure 1: Proposed Authentication Block Diagram

While STFT gives a constant resolution at all frequencies, the Wavelet Transform uses multi-resolution technique by which different frequencies are analyzed with different resolutions. In above block diagram the input we are taking is a CT scan image and MRI images in the form of text file. For the text file we convert the actual images using MATLAB GUI interface. The Wavelet Series is just a sampled version of CWT and its computation may consume a significant amount of time and resources, depending on the resolution required [5]. The Discrete Wavelet Transform (DWT), which is based on sub-band coding, is found to yield a fast

computation of Wavelet Transform. It is easy to implement and reduces the computation time and resources required. In CWT, the signals are analyzed using a set of basis functions which relate to each other by simple scaling and translation. In the case of DWT, a time-scale representation of the digital signal is obtained using digital filtering. A) DWT Image Fusion The Discrete Wavelet Transform was developed to apply the wavelet transform to the digital world and Filter banks are used to approximate the behavior of the continuous wavelet transform. The signal is decomposed using a high-pass filter and a low-pass filter. The coefficients of these filters are computed by doing mathematical analysis and available in subsections 1. First register the input images (I1 and I2), which are going to be fused and corresponding pixels are aligned. 2. The Registered input images are decomposed into wavelet transformed images respectively, based on haar wavelet transformation (W). The transformed images with K-level decomposition will include one low-frequency portion (LL band) and 3K high-frequency portions (LH bands, HL bands and HH bands). 3. The Transform coefficients of different portions are performed with a certain fusion rule. 4. Then by applying performing an inverse wavelet transform (W-1) based on the combined transform coefficients, the fused image (I) is constructed

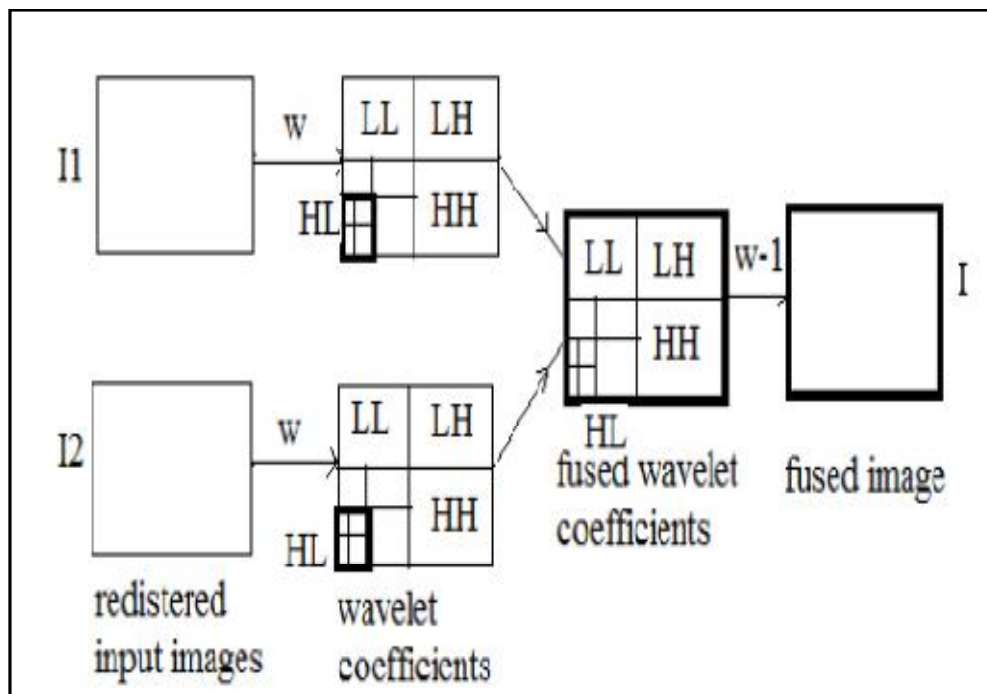


Figure 2: The Scheme of image fusion using wavelet transforms

3.1. DWT Decomposition of Images

Filter banks decompose the signal into high- and low-frequency components. The low-frequency component usually contains most of the frequency of the signal. This is called the approximation. The high-frequency component contains the details of the signal. Wavelet decomposition can be implemented using a two-channel filter bank. In wavelet decomposing of an image, the decomposition is done row by row and then column by column. For instance, here is the procedure for an $N \times M$ image. Then filter each row and down-sample to obtain two $N \times (M/2)$ images, they are L and H images. The formulas used to do find H and L is given below. $H = \text{odd-even} = F1$; $L = \text{even} + \text{round}(F1/2)$; Then filter each column and sub sample the filter output to obtain four $(N/2) \times (M/2)$ images, by using the formulas shown in below. $LH = \text{odd} - \text{even} = F2$; $LL = \text{even} + \text{round}(F2/2)$; $HH = \text{odd} - \text{even} = F3$; $HL = \text{even} + \text{round}(F3/2)$; The four sub-images obtained as seen in Fig. 1, the one obtained by low-pass filtering the rows and columns is referred to as the LL image. Similarly LH, HL and HH are also formed. Each of the sub images obtained in this fashion can then be filtered and sub-sample to obtain four more sub images. This process can be continued until the desired sub band structure is obtained. Here F1, F2 and F3 are taken as references.

4. Implementation Process

The implementation steps of the image fusion using DWT is done as follows. MATLAB is used to convert input images into header files using GUI feature. [3]. Xilinx Platform Studio (XPS) is a part of an EDK system and it includes the XPS-GUI and all tools run by the GUI to process hardware & software components within XPS.

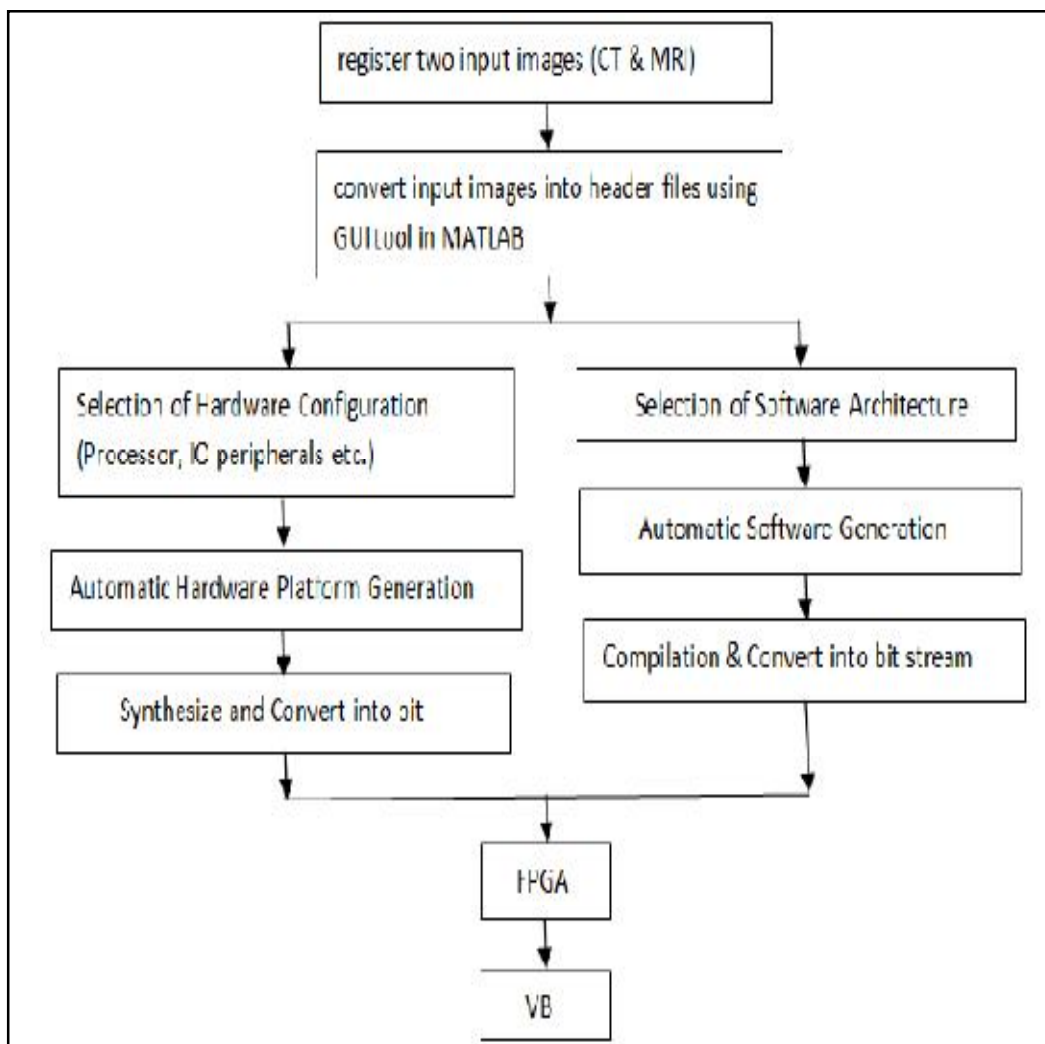


Figure 3: Flow chart for FPGA implementation of image fusion

FPGA hardware is chosen because of it has some advantages like reconfigurability, low power dissipation, small size [7]. The process is done as per below. Automatically for the each specified hardware selection and then convert it into bit streams. Step 1: First Select the Hardware configuration like processor, IO peripherals, etc. The HDL codes are generated

Step 2: After that the selection of software architecture is started. It selects drivers, libraries, etc. the header files of input images and source file (which consist of DWT, fusion and IDWT code written in the system-C language) are added and convert it into the bit stream. Step 3: Then download hardware and software architecture bit streams into FPGA and run it. The image output can't be shown in FPGA. For this purpose Visual Basic is used to observe the fused image output.

5. Result and Analysis

5.1. Input

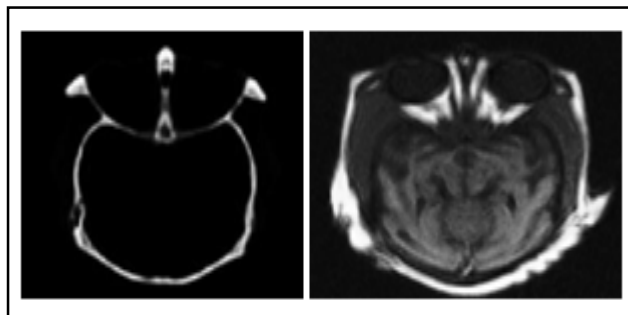


Figure 4: I/P of CT scan Image and MRI Scan Image In .Jpg Format

5.2. Output of Existing Technique

5.2.1. Output of Simple Minimum Technique

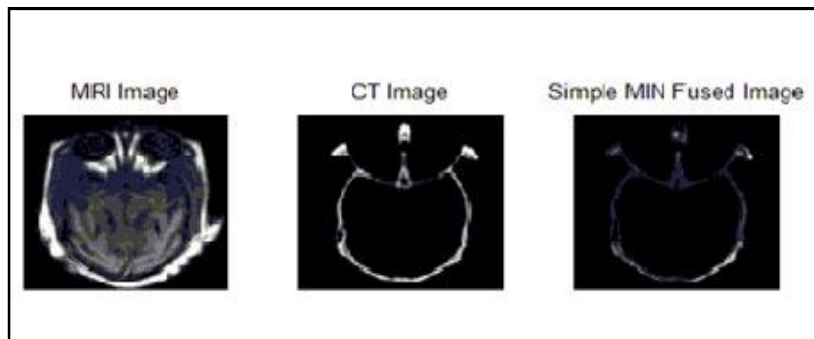


Figure 5: O/P of Simple Minimum Technique

5.2.2. Output of Simple Maximum Technique

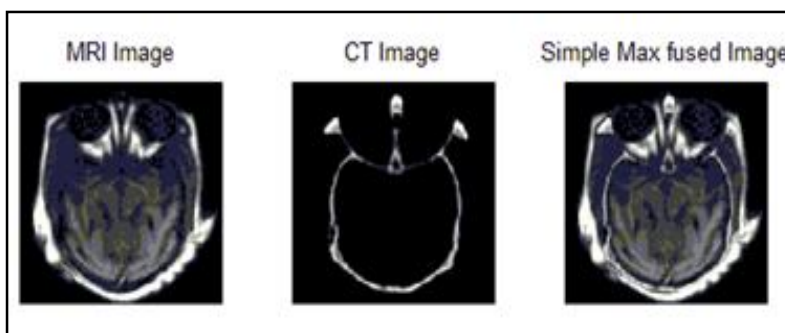


Figure 6: O/P of Simple Maximum Technique

5.2.3. Output of Simple Average Technique

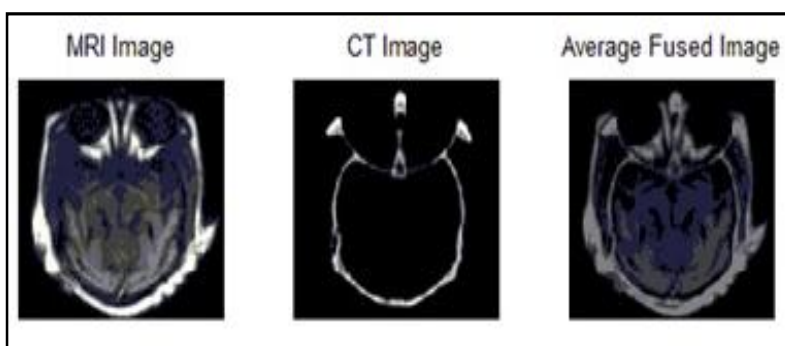


Figure 7: Image of O/P of Simple Average Technique

5.3. Output of Simple PCA Technique

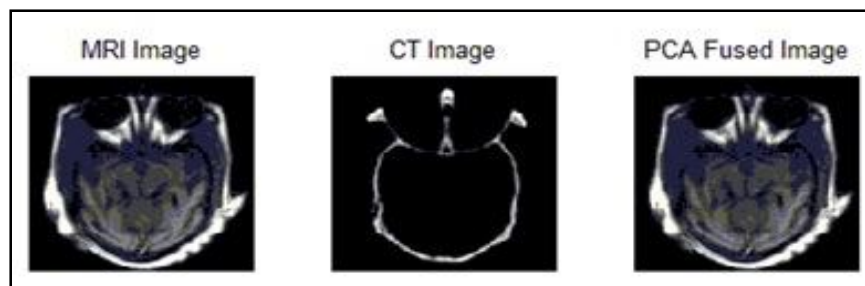


Figure 8: O/P of PCA Technique

5.3.1. Output of Proposed System

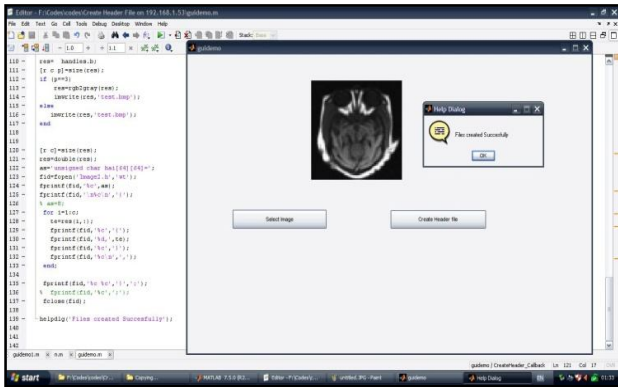


Figure 9: I/P image 1 for MATLAB

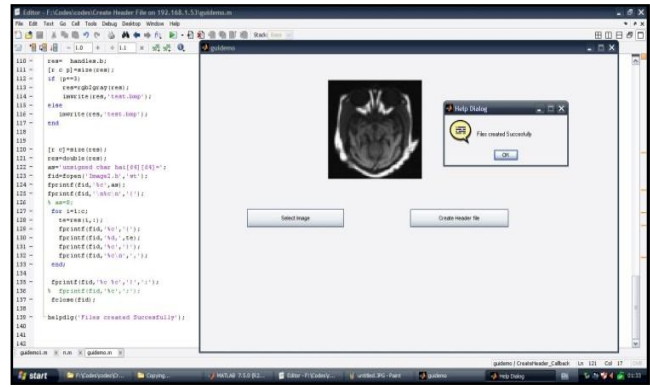


Figure 10: Input Image 2 for MATLAB

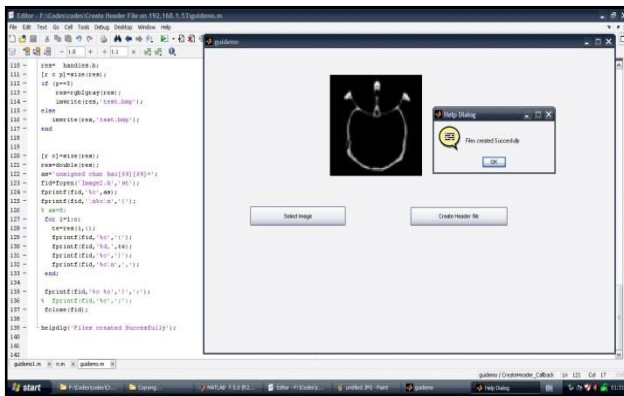


Figure 11: Input Image 1 for MATLAB

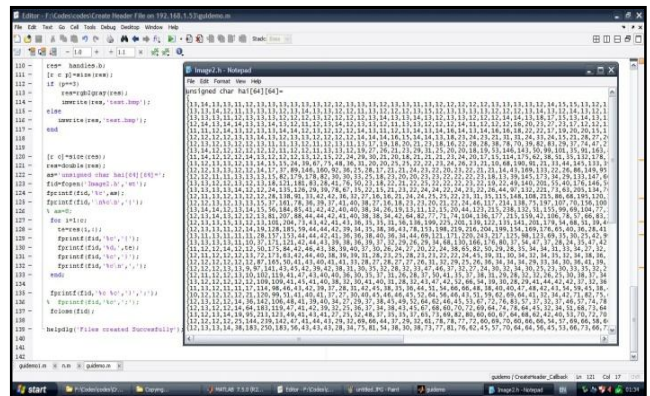


Figure 12: Image 1 Header

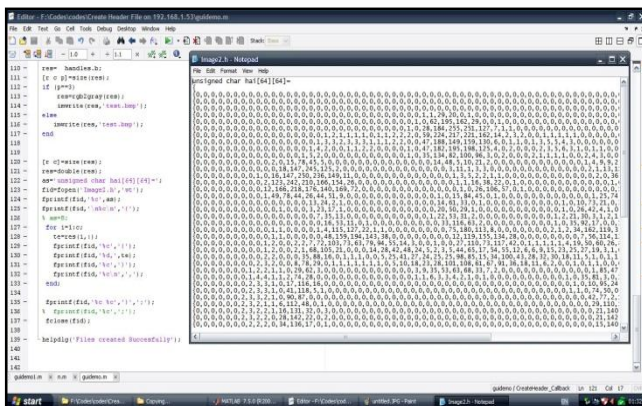


Figure 13: Image 2 Header

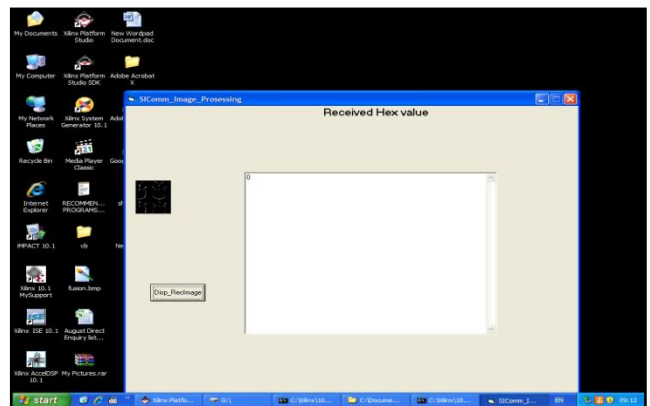


Figure 14: DWT of Image 1

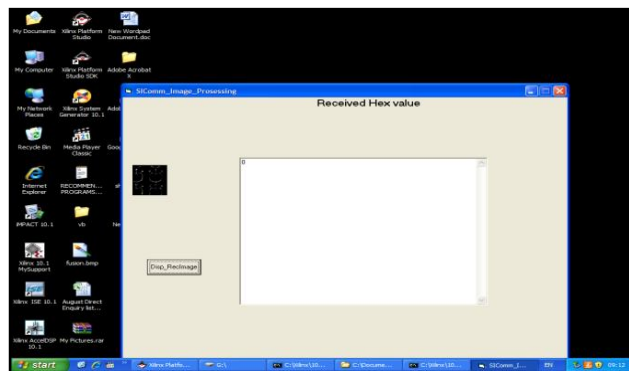


Figure 15: DWT of Image 2

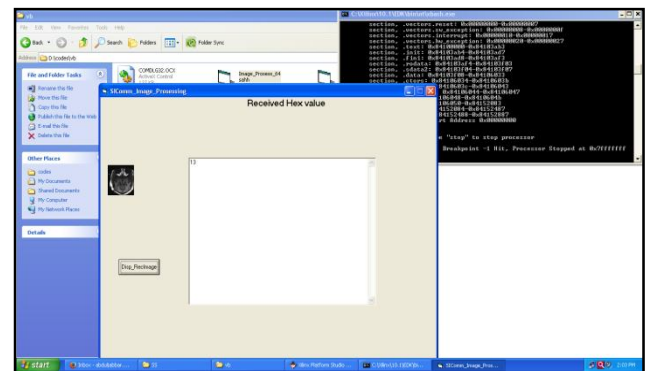


Figure 16: Fused Image Using DWT

5.4. Fusion Techniques & Their Quality Measures

Fusion Method	Quality Measurements	
	Peak Signal to Noise Ratio	Mean Square Error
Simple Minimum	10.5386	574.3
Simple Maximum	19.9507	657.2
Simple Average	16.0485	161
PCA	35.23	19.48
DWT	43	3

Table 5.2: Quality Measurement of Image Fusion

5.4.1. Peak Signal to Noise Ratio

The PSNR is use to calculate the similarity between two images. The PSNR between the reference image R and the fused image F is defined as. For better fused image PSNR value should high.

$$\text{PSNR} = 10 \log_{10}(\text{Peak}^2 / \text{MSE})$$

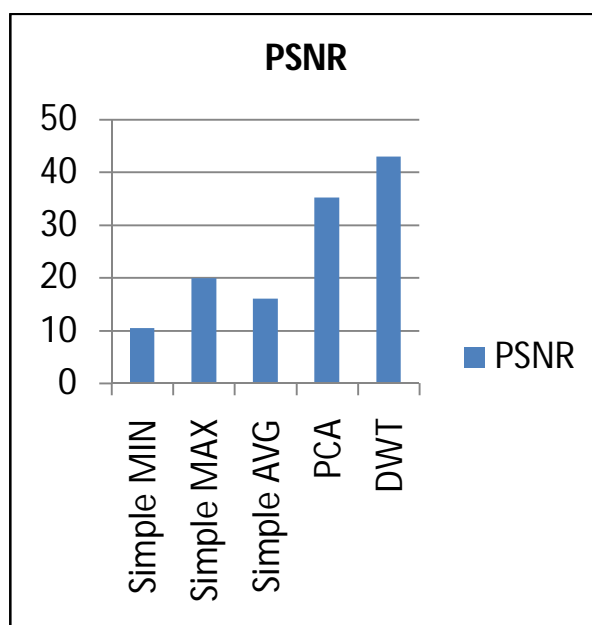


Figure 17: Graph Showing Comparison of Different Technique Based On PSNR

5.4.2. Mean Square Error

Mean square error measures the average of the square of the error which gives the difference between fused image and source image. MSE should be as low as possible for better result [10].

$$\text{MSE} = \frac{1}{mn} \sum_{i=1}^m \sum_{j=1}^n (A_{ij} - B_{ij})^2$$

5.5. Advantages

1. Fusing X-ray computed tomography (CT) and magnetic resonance (MR) images
2. Useful in a Computer assisted surgery
3. Spatial registration of 3-D surface
4. Its advantageous in Electronic circuit and component inspection
5. Used in Product surface measurement and inspection
6. Intelligent robots on assembly lines
7. Complex machine/device diagnostics [9].

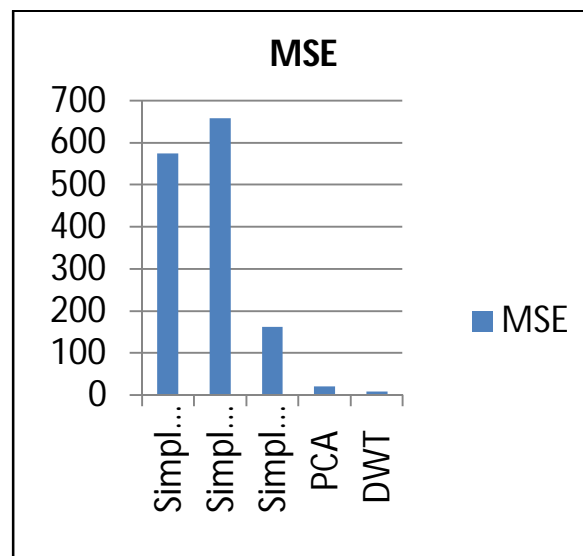


Figure 18: Graph showing comparison of different technique based on MSE

6. Conclusion

In this paper, the hardware implementation of a fusion method that is suitable for remote sensing data has been presented. The DWT fusion method is configurable since it allows the user to control the correlation properties of the final fused color image. The hardware realization which is based on FPGA technology provides a fast, compact, and low-power solution for image fusion. The dedicated sections provide a detailed description of the methodology to transform the DWT fusion method in a hardware realizable process.

7. Acknowledgements

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