

Multi-modal medical image analysis using Wavelet Fusion

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Abstract - Techniques for pixel-level image fusion have been the most important for remote sensing data processing and analysis up until this point. Typically based on empirical or heuristic rules, feature based fusion techniques are utilized for this purpose. Multimodal transport image registration and fusion technologies play an important role in routine screening, screening, screening and evaluation of chronic disease radiotherapy, surgical and radiotherapy programmes. Multimedia media algorithms and tools have made great strides in supporting the reliability of clinical decisions on medical imaging and will continue to make great strides. Combining the two types of information and mixing the two images. Image aggregation methods include simple methods (e.g. pixels) and complex methods (such as wavelet transforms). The advantage of using wavelet manipulation is it has a large part of each image. Its main objective is to improve the understanding of medical images through the use of discrete wavelet transformation technology. DWT uses mainly consolidation rules involving average pixels. The discrete wavelet transformation was carried out using fusion techniques designed specifically for integrated medical images. The fusion performance is calculated based on PSNR, MSE and whole progression moment.

Keywords- Multimodal medical image fusion, fusion rules. , DWT,HAAR,MRI,PET

I. INTRODUCTION

Typically, the word "fusion" is followed by a number of other words, such as "merging," "combination," "synergy," and "integration," all of which roughly convey the same meaning that the concept has had ever since it first appeared in literature. Each author interprets this term in a different way based on his or her research interests, as in [1,2].The following is a general definition of data fusion: A formal framework for data fusion describes methods and tools for combining data from various sources. It aims to acquire higher-quality information; The combination of images means that the process of reconciling a large amount of image information with the new final image is appropriate and consistent with the ultimate goal of human visual recognition and computer processing.

The combined images should gradually have full data, which are increasingly valuable for human or mechanical observations. The advantage of a combination of images [3][4] is increased reliability and reliability. In medical numbers, CTs, attractive reverse imaging MRIs, PETs, SPECTs and field image processing methods that reflect human data at different locations. In the article, we will present and discuss the objectives of combining the computed tomography images and magnetic resonance images To date, a number of procedures have been planned for the consolidation of written images. Some technologies are identified through the integration of multimodal picture. Picture fusion mainly controls three types, such as pixels, characteristics and decision level

techniques. Because of its ease of implementation and computing capacity, pixel-grade photo synthesis systems are often used to mix medical images. There are a number of conditions necessary for the process of picture fusion process, [5] the consolidation of the picture must retain all the important sequences entered into the picture combination, but cannot start the object that could lead to a misdiagnosis. First of all, incredible and significant steps must be taken.

The corresponding characteristics, photo tabulation methods may be divided into three categories: point-based, external or scale-based, [6][5][9]. The point-based check requires the coordinates of the relevant position in the special picture, and these corresponding points are used to estimate geometry conversions. External support checks require the identification of image surfaces to match and minimize spatial measurements on this generating plane. The base registration requires optimal consideration of data from certain predefined features that measure the matching of all geometric results. Projection integration methods are rare and the average of non-efforts using SNR (signal/noise ratio) [10]. Fusion methods that operate on such transform-domain features have been shown to produce subjectively superior fused images to pixel-based methods [11].

II. RELATED WORK

In medicine, image-processing technologies play an important role. Electronic data processing automation is

the most real and prominent method. Brain diseases can be identified by magnetic characteristics (magnetic resonance) and (PT). Additional magnetic resonance and PET scan variants were used for medical diagnosis. Medical experts need solid computational examinations and their related analysis. To support disease diagnosis, accurate information under various remediation image methods. i.e. CT, PT and MRI).

The image combination is a technique that integrates two images into one image. Image fusion may occur at the following processing stage with feature-level methods. The extraction of features from the input images is necessary for feature-level fusion. Depending on the intended use of the fused image and the nature of the images, various types of features are taken into consideration. Pixel intensities or edge and texture features are examples of features [11]. The extraction of feature primitives like edges, regions, shape, size, length, or image segments, as well as features with similar intensity in the images that will be fused from various types of images.

Compared to the original single-scan image, the single fusion image obtained using several multimodal medical images is an improved anatomy very desirable spectral information. This image obtained using multimodal fusion picture is positive for experimental analysis by skilled health-care personnel. Inspection work; structured to provide magnetic resonance imaging or PET. Pre-processing techniques remove noise from images, resize images and filter into images for filtering, and we apply a Gaussian filtering technique with spatial filter technology. Advanced images are transmitted through (DWT) to different areas of the head image. This structure achieves approximately 90-95% of perfect results through moderate change. The acquisition of Fusion images obtained does not include the presence of ethers or anatomy records. An experimental examination of an image data set of Alzheimer's disease, mean attribution or conventional coronary syndrome was carried out. Measurements and graphs are shown in DWTs, which can significantly improve the combination of features.

Fusion of medical images: Multimodal medical imaging algorithms and equipment show remarkable discoveries in the evolution of correctness in medical imaging decisions. The selection of imaging models for objective clinical research requires the specific medical knowledge of research institutions. It is almost impossible to imprison the imaging factors of the imaging model to ensure that the analysis and analysis of the results is medically accurate. (a) Identification, improvement and development of useful imaging methods for the concentration of medical fusion (b) development of different techniques for the integration of medical fusion technologies (c) application of techniques for the integration of medical fusion planned for the study of concentrated human organs and the assessment of medical conditions.

III. PROPOSED APPOROCH

It is proposed a medical image fusion algorithm based on wavelet transform algorithm. The algorithm achieves PET/CT fusion through wave coefficient fusion. The innovative thought in addition to premise of multi-solution testing stand on wavelet comes beginning Mallat. The Wavelet transformation is an arithmetical implement so as to be able to notice local characteristics in the indication procedure. It can also be used to separate the resolution from the two-dimensional (2D) signal (e.g., a 2D greyscale image display) for multiple resolution analysis. Wavelet conversion is widely used in many areas, as it is related to texture analysis, data compression, feature detection and image fusion. In this paragraph, we temporarily re-evaluate and analyse wavelet image fusion techniques.

$$(x, y) = W^{-1}(\emptyset(W(I_1(x,y)), (W(I_1(x,y))))$$

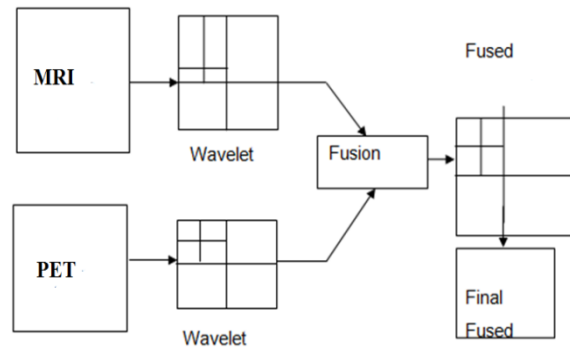


Figure1: Image fusion using DWT.

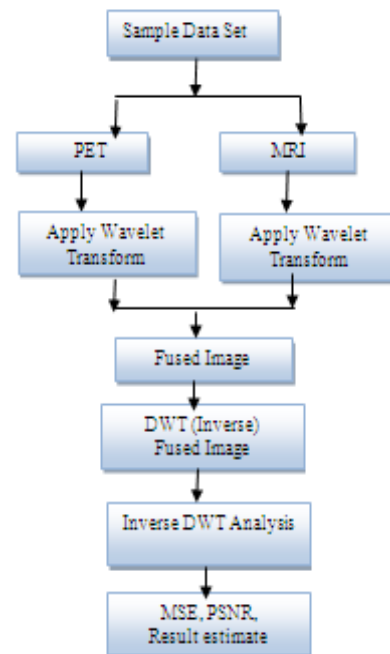


Fig.2 Proposed Flow Chart

Wavelet Transformation Wavelet study is in line with the next logical step; window technology, with different sizes. Longer time intervals can be used when more precise low frequency sequences are needed, and shorter time intervals can be used in areas where high frequency in sequence is required. In individual measurement (1D), the underlying idea of DWT is to indicate overlap of waves. Assuming that a separate indication represents through $f(t)$; then, the wavelet is determined to be $(t) = m, cm, \psi_{m,n}(t)$ Where $\psi_{m,n}(t) = 2^{-m/2} [2^{-m} t - n]$ along with m and n are integers.

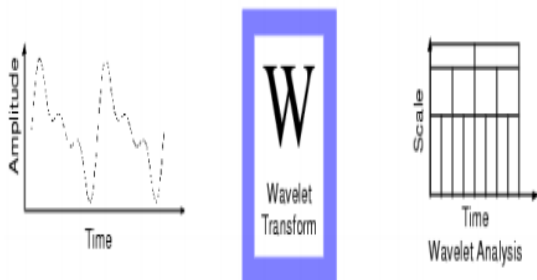


Figure 3: Discrete wavelet transform

Algorithm- has been developed and implemented using MATLAB software.

STEPS:

1. Read and find the size the image I1.
2. Read and find the size the image I2.
3. Calculate size, if not identical, match size, make size equal
4. Convert above images from grayscale to indexed image to perform various wavelet functions. If the color map is horizontal, the wavelet convert can be openly applied to the indexed representation otherwise the indexed image should be converted to grayscale formatting
5. Using any wavelet (haar,) perform multilevel wavelet decomposition.
6. Generate a matrix of coefficient of the level-III estimate and straight, perpendicular and oblique information.
7. From the coefficients, construct and display approximations and details.
8. By multilevel inverse wavelet transform, regenerate an image.
9. Repeat the above steps with second image.
10. Now integrate or fuse the wavelet coefficients using either of averaging, maximum or minimum technique.
11. Generate a final matrix of fused wavelet coefficients.
12. Calculate the inverse wavelet transforms so as to obtain the fused image.
13. Now calculate the PSNR and MSE.
14. Display the results.

Results- We considered wavelets namely Haar, for fusing the two input images are first read and then converted into PET and MRI images. There are rules for fusion. This is because, in the case of PSNR and MSE haar wavelet along with maximum rule produced better results and is therefore used for further analysis.

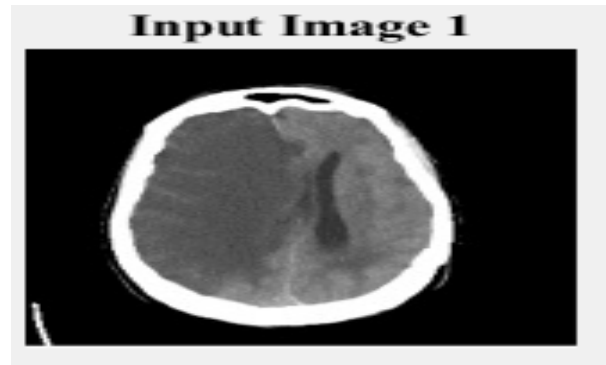


Fig. 4 MRI image

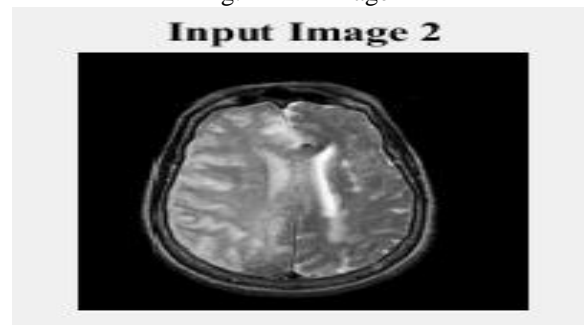


Fig. 5 PET image

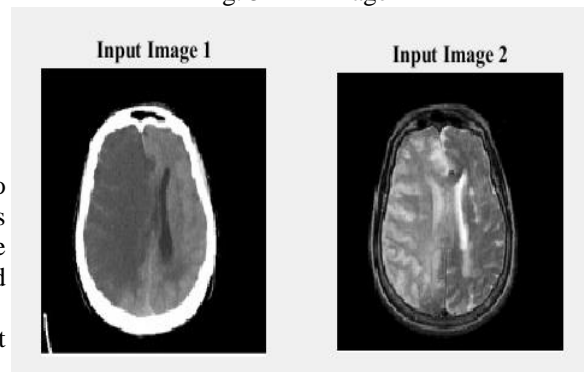


Fig. 5 Input images (MRI & PET)

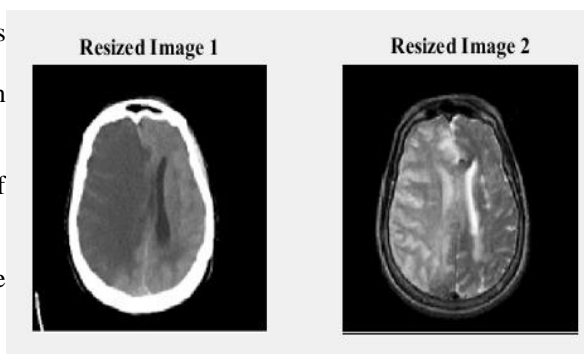


Fig. 6 Resized MRI & PET image .

The coefficients found be afterward fused used a specific combination rule and then restores back the image using inverse discrete wavelet transform.

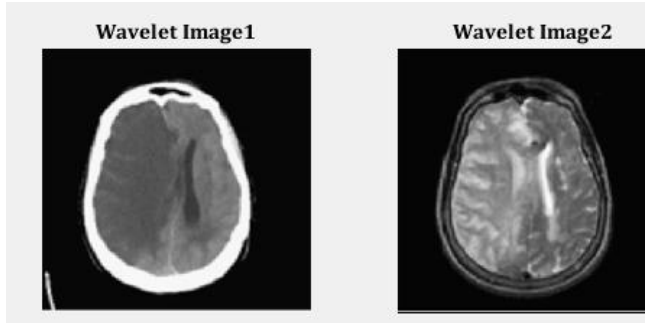


Fig. 7 Wavelet Transform of MRI&PET images.

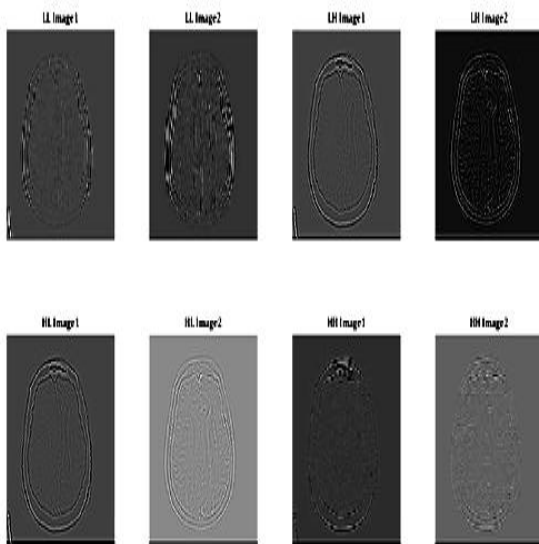


Fig. 8 Wavelet Transform coefficient of MRI & PET images .

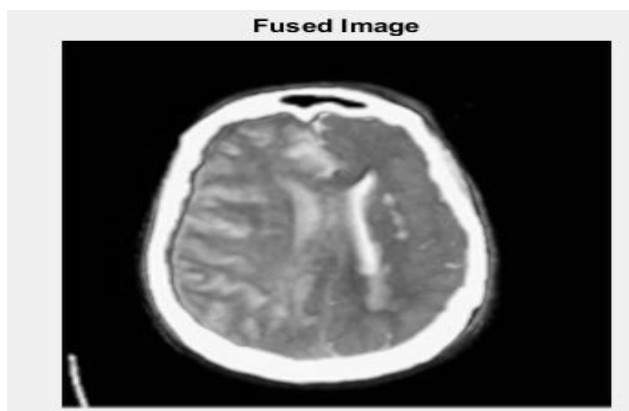


Fig.9. Fused image

A graphical presentation of the system developed internally by MATLAB is presented in outline 10. In this system, Haar wavelets are used for the fusion method.

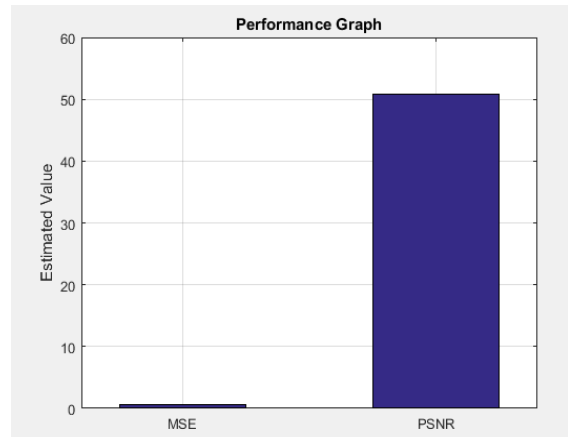


Fig 10 MSE and PSNR Performance graph

Table I. Performance Measure Based On PSNR, MSE of the Output Fused Image

Performance	Previous work	Proposed work
PSNR	41.32	48.10
MSE	0.011	0.011

IV. CONCLUSION

The integration of disparate and complementary data is the goal of Image Fusion, which aims to improve the information in the images and make interpretations more reliable. This results in data that is more useful and more accurate. This article proposes a wave fusion method for magnetic resonance and PET images. The wavelet decomposition of the data set will be divided into four levels, with low and high activity areas. This experiment will test the haar wavelet method. Haar's wavelet was used to consolidate the database of multiple 3D medical models. To test the quality of the merged image, pre-processing will apply a Gaussian filter (spatial filtering), using MSE and PSNR.

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