SHIFT INVARIANT AND EIGEN FEATURE BASED IMAGE FUSION

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ABSTRACT:

Image fusion is a technique of fusing multiple images for better information and more accurate image compared input images. Image fusion has applications in biomedical imaging, remote sensing, pattern recognition, multi-focus image integration, and modern military. The proposed methodology uses benefits of Stationary Wavelet Transform (SWT) and Principal Component Analysis (PCA) to fuse the two images. The obtained results are compared with exiting methodologies and shows robustness in terms of entropy, Peak Signal to Noise Ratio (PSNR) and standard deviation.

KEY WORDS:

Fusion, multi-focus image integration, SWT, PCA.

1. INTRODUCTION

Image fusion is for obtaining more accurate information by combination of two or more images. The fusion has many applications like biomedical imaging, air traffic control, robotics, remote sensing, pattern recognition, multi-focus image integration, and modern military environmental monitoring [1].

The image fusion can be achieved in spatial domain or frequency domain. Jiayi Ma et al., proposed a technique of fusing infrared image and visible image. The fusion of two images is depended on Gradient Transfer Fusion (GTF), GTF is a combination of gradient transfer with minimization of total variation. Jiayi Ma et al., are succeeded image fusion without any registration of source images [5]. Yanfei and Nong [6] proposed a multi sensor image fusion technique based on hierarchical multi resolution along with attention. Important areas are identified by using visual attention model and maximum entropy. Based on adoptive weighing rules, first level of fused image is obtained from visible image and infrared image. Finally Non Subsampled Counterlet Transform (NSCT) is used to obtain final fused image. Huafeng Li [7] also proposed for the fusion of multi sensor image combination based on NSCT. Jun Lang and Zhengchao image fusion technique [8] provides less spectral distortion and good spatial resolution based on discrete fractional random transform and adaptive pulse coupled neural network (PCNN).

Principal Component Analysis (PCA) [2] increases the spatial resolution and the flaw is creation of distortion in spectrum in remote sensing image fusion. SWT advantage translation invariance over DWT and DWT advantage is time frequency localization. The paper set as follows, in section 2 proposed methodologies, its flow diagram and information about techniques used.
presented. In section 3, results and performance calculations and analysis are presented. Finally conclusions along with future directions are given in section 4.

2. **PROPOSED METHOD**

The proposed technique uses the advantage of SWT and PCA. Initially, the input images of any size are registered into same size, and then given to SWT separately. The SWT converts an image into four different subbands A, H, V and D. In figure1 A1 and A2, H1 and H2, V1 and V2, D1 and D2 are approximate coefficients, horizontal details, vertical details, diagonal details of image1 and image2 respectively. The approximate coefficients are fused based on PCA, similarly horizontal details, vertical details and diagonal details also fused individually. After fusion of respective coefficients, the fused coefficients are given to ISWT, which results final required fused image.

2.1 **Stationary Wavelet Transform**

The drawback in discrete wavelet transform is translation invariance is overcomes in stationary wavelet transform. Due to presence of up-samplers and down-samplers in DWT, it lacks translation invariance and this effect is eliminated in SWT by removing the up-samplers and down-samplers, so SWT also called as Translation invariant wavelet transform. The outcome of SWT consist same number of samples as the input, so it is also called redundant wavelet transform.
2.2 Principal Component Analysis

PCA is a statistical procedure based on orthogonal transformation; to convert set of structure of possible correlated variables into set of linearly uncorrelated variable called “principle components”. Output after PCA consists of less number of Eigen features when compared to original input spectral feature. In this method initially need the mean values which is given in Equation (1) of spectral features later calculate the covariance matrix in Equation (2), based on this covariance matrix find the feature vectors. To find out the Eigen features by using singular value decomposition (SVD) method in Equation (3).

\[
E(x) = \frac{1}{k} \sum_{i=1}^{k} x_i \tag{1}
\]

\[
\text{Cov}(X) = (X_i - \mu) (X_i - \mu)^T \tag{2}
\]

\[
A = U \Sigma V^T \tag{3}
\]

![Flow diagram of PCA](image)

1. Results

The results of proposed methodology are displayed in below tables and here the testing database downloaded from standard organizations [9], [10]. Table 1 displays, proposed method results from dataset 1 [10]. Table 2 displays, proposed method results from dataset 2 [9]. All the process done on personal computer with RAM 2GB, Matlab version 2013.
Table 1: Comparison of Fused results for dataset 1.

| Image/Measurement | Entropy PCA | Entropy SWT | Entropy PCA+SWT | Standard Deviation PCA | Standard Deviation SWT | Standard Deviation PCA+SWT |
|-------------------|-------------|-------------|----------------|-------------------------|------------------------|---------------------------|
| Pair1(DU)         | 6.25        | 6.99        | 7.40           | 25.62                   | 25.98                  | 29.73                     |
| Pair2(NU)         | 6.08        | 6.59        | 7.43           | 11.99                   | 13.32                  | 15.70                     |
| Pair3(FLY)        | 4.66        | 4.65        | 5.68           | 41.87                   | 42.19                  | 41.89                     |

Table 2: Comparison of Fused results for dataset 2.
Performance Calculations

For testing of proposed methodology, PSNR, Standard Deviation and Entropy are used [4], [3]. The performance calculations are presented in below table3 and table4. PSNR1 represents for peak signal to noise ratio of fused image with image1 and PSNR2 for peak signal to noise ratio with image2.

| Image\Measurement | PSNR 1 | Fused image by | PSNR 2 | Fused image by |
|-------------------|--------|----------------|--------|----------------|
|                   | PCA    | SWT            | PCA    | SWT            |
| Pair1(DU)         | 23.00  | 22.86          | 21.20  | 21.19          |
| Pair2(NU)         | 20.84  | 27.00          | 39.27  | 23.29          |
| Pair3(FLY)        | 26.65  | 24.19          | 38.45  | 26.70          |

Table3: Performance calculations comparison in PSNR.

Table4: Performance calculations comparison in Entropy and Standard Deviation.

CONCLUSIONS

The proposed fusion methodology is done by the help of SWT and PCA, SWT has the advantage of shift invariance over Discrete Wavelet Transform. The comparative analysis of performance calculations showing robustness over remaining two techniques. The future idea is to fuse infrared images.
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