

# Region-based Image Fusion Method with Dual-Tree Complex Wavelet Transform

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**Abstract.** Dual-Tree Complex Wavelet Transform is an enhancement to the discrete wavelet transform, with important additional properties: it is nearly shift invariant and directionally selective in two or higher dimensions. And also can overcome the problem of distortion in spectral and spatial resolution of fused image with other techniques. Combined with region-based fusion method, it can obtain better fusion results. All these had been proved by the experiment.

**Keywords:** Dual-Tree Complex Wavelet Transform; Region-based fusion; Shift invariant; Resolution; Directionally selective.

## 1. Introduction

Multispectral sensors are increasingly being employed in satellite imagery of the earth, multispectral data is required in order to extract the maximum amount of information from a scene, provide the desired fusion image of information into a single picture for rapid assessment [1-4]. All these rely on the image fusion, the combination of the multi-sensor images fully take into account the complementary and supplementary provided by different data sources and optimize the classification of cartographic objects. Thus many image fusion techniques and software have been developed, such as the well-known Principle Component Analyses (PCA), Intensity Hue Saturation (HIS), and Wavelet-based image fusion and so on. One drawback of PCA and HIS is some distortion of spectral characteristics in the original multispectral image [2]. Wavelet-based image fusion can overcome this problem, and provides high spectral quality of the fused satellite images [5], however, it reduces much spatial information, so an improved method has been proposed in this paper that can keep the spectral and spatial resolution that is Dual-Tree Complex Wavelet Transform (DT-CWT).

## 2. Region-based Image Fusion Method with Dual-Tree Complex Wavelet Transform

### 2.1 Region-based image fusion with complex wavelets

DT-CWT is an enhancement to the discrete wavelet transform, with important additional properties: it is nearly shift invariant and directionally selective in two or higher dimensions. The multidimensional DT-CWT is non-separable but is based on a computationally efficient separable filter bank.

The region-based fusion can obtain the best fusion results by considering the nature of points in each region altogether, and use some segmentation algorithm to separate an original image into different regions, and then design different rules for different regions[6-9]. For 2-D image, it can be filtered separately along columns and then rows by the way like 1-D. To represent fully a real 2-D signal, it must filter with complex conjugates of the column and row filter. Furthermore, it remains computationally efficient, since in fact it is close to a classical real 2-D. Wavelet transform at each scale in one tree, and the discrete transform can be implemented by a ladder filter structure[3]. It can reserve the needed details or the approximation of a given scale, while remove all other scales, and shift the input image to produce a shift of the reconstructed filtered image without aliasing. The superiority of DT-CWT is that it can separate more directions than the real wavelet transform. The 2-D DWT produces 3 band-pass subimages at each level, while the 2-D DT-CWT can provide 6 band-pass

subimages in two adjacent spectral quadrants at each level, so the positive frequencies can be separate from negative ones vertically and horizontally. Figure 2 shows the region-based fusion method for infrared and visible images adapts the DT-CWT for its nearly shift invariant and limited redundancy. The diagram of the proposed region-based method with DT-CWT as Figure 1 showing:

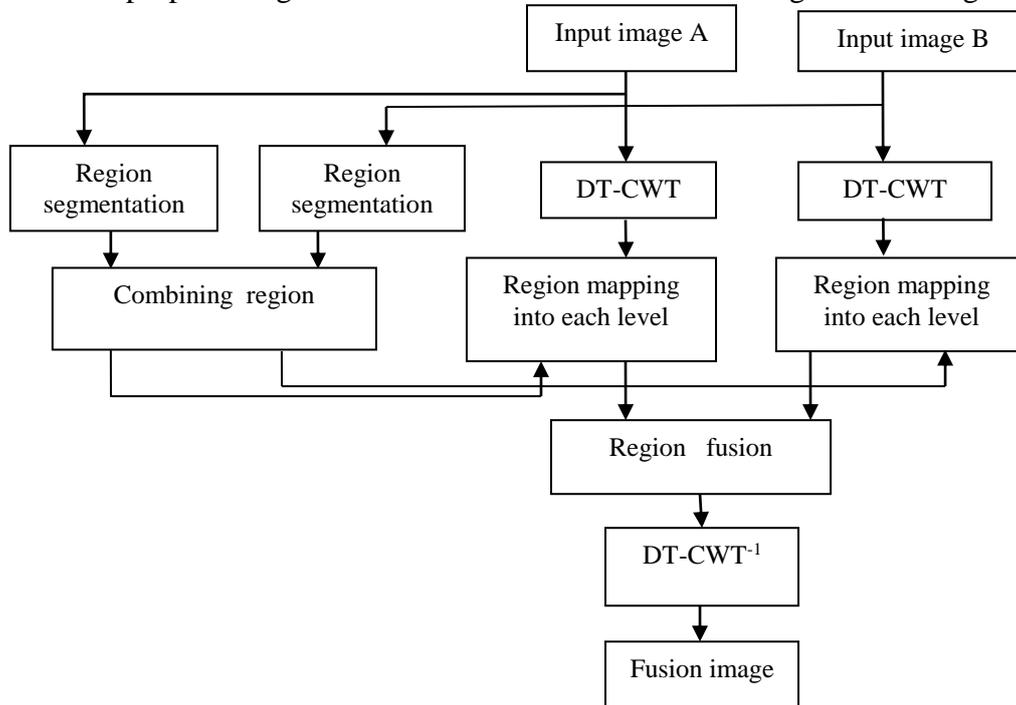


Figure 1 The diagram of the proposed region-based method with DT-CWT

## 2.2 Image fusion based on region segmentation and complex wavelet

As Figure 1 showing, Segmentation is firstly performed on the low level visible light image marked A and infrared image marked B respectively, consequently, the DT-CWT coefficients from the different regions are merged separately, finally the fused image is obtained by operating inverse DT-CWT.

Decomposed by the multi-resolution DT-CWT, low-frequency part of the images denotes their approximate components, which contains spectral information of the source image. High-frequency part of the images denotes their detail components, which contains edge detail information of the source image. Usually the fusion rules are commonly that average operator is used in low frequency domain, and max absolute operator is used in high frequency domain. For the two fused source images of the same scene, spectral information of visible light image is much richer than the infrared image. When the fusion of average operator is adopted, part of spectrum information of visible light image will be lost. In order to settle this problem, we adopt spatial frequency to guide region-based fusion. The spatial frequency has led to an effective objective quality index for image fusion.

## 2.3 Image fusion experiment

An experiment has been done in the method of the above, and the result showed in Figure 2. The infrared and visible image of the same scene are fused by region-based method with DT-CWT. Meanwhile region-based colour fusion is used, that is the image segmentation is firstly carried out, then region colouring is realized. Its major points are (a) the infrared and visible images are rendered segment-by-segment; (b) the segmented regions of the two images are combined and form a new segmented map; (c) the regions are classified according to the target types and the spatial frequencies. According to the fusion images, this method has good effectiveness, which preserves not only the spectral information of the visible light image, but also the thermal target information of the infrared image. According the fused image in Figure 2, it should be noted that both the spatial and spectral resolutions have been enhanced, in comparison to the original images. The spectral information has been increased and the structural information in the original multispectral image has also been enriched, so the fused image has a better visual effect.

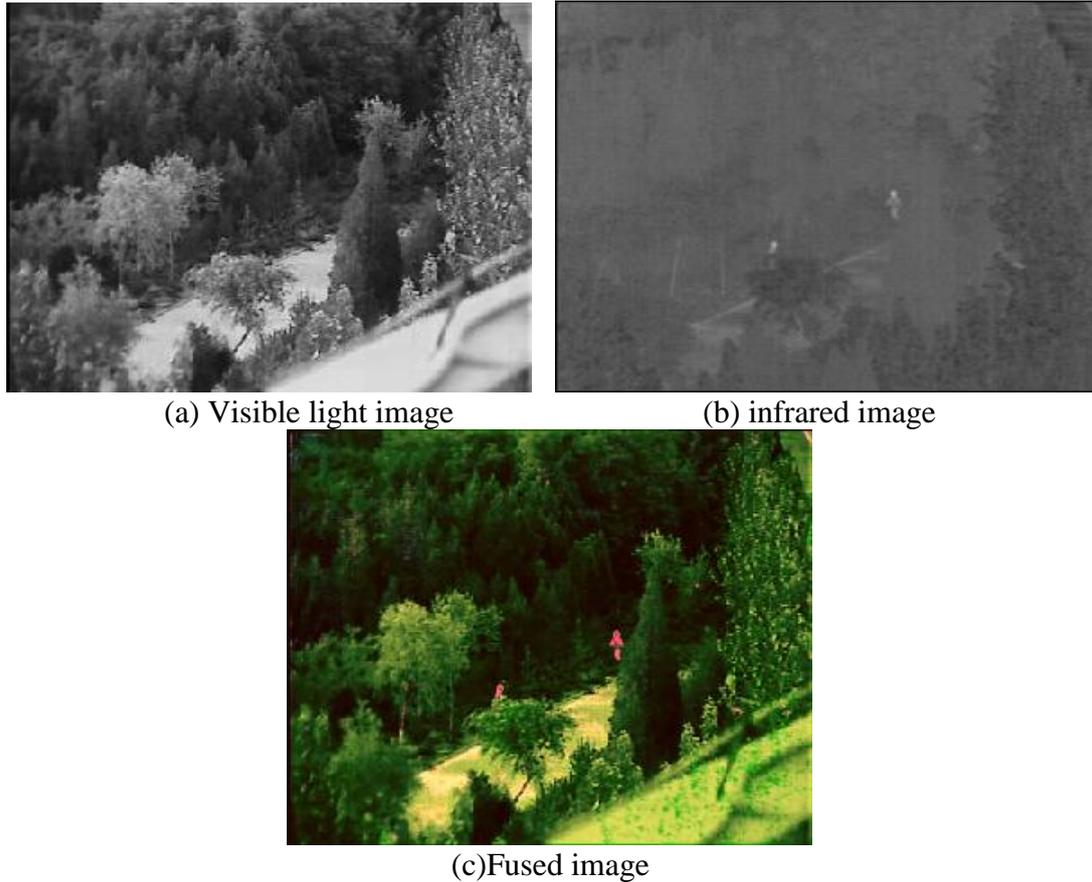
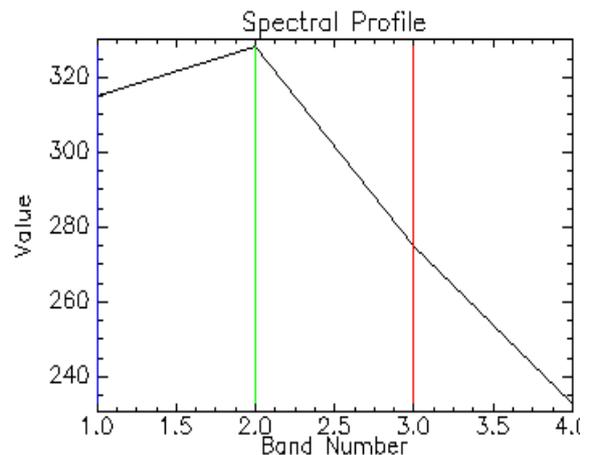
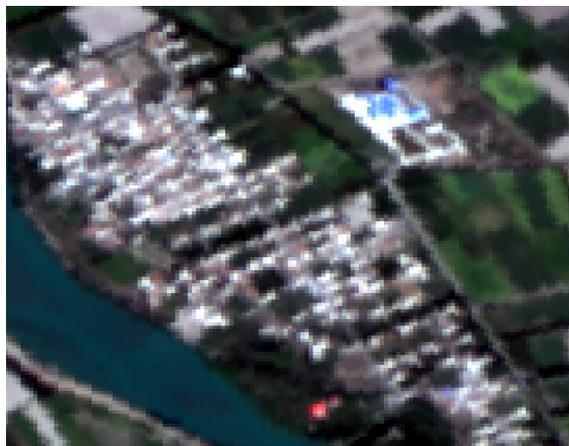


Figure 2. Visible light image and infrared image fused by region-based method with DT-CWT

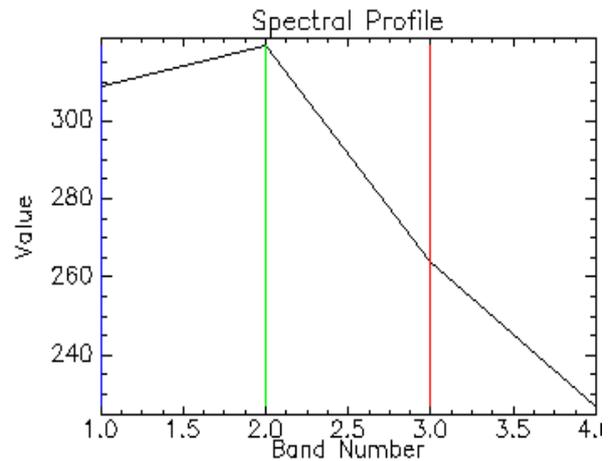
Since the human eye can discern several thousand colours, while it can only distinguish about 100 shades of gray in any case, so mapping multiple spectral bands of imagery into a 3-D colour space is presented, which increase the dynamic range of a sensor system. Experiments have convincingly demonstrated that appropriately designed false colours rendering of imagery can significantly improve observer performance and reaction times in tasks that involve scene segmentation and classification. However, inappropriate colour mappings may hinder situational awareness, as Figure 3(a) Showing. A solution to this problem is ergonomic colour scheme that can produce imagery with natural appearance and with colours invariant for change in the environmental conditions, as Figure 3(c) Showing. Compared Figure 3(a) and (c), it is clear that in Figure 3(c) the counter intuitive appearance of the scene is more legible.



(a) 3-D colour space mapping by false colour (b) The mapping multiple spectral bands of imagery



(c) Image fusion with ergonomic colour scheme



(d) Ergonomic colour scheme bands

Figure3. Image fusion with 3-D colour space mapping and ergonomic colour scheme

### 3. Conclusion

The proposed method which uses Dual-Tree Complex Wavelet Transform and region-based method is applied to fuse the infrared and visible image of the same scene, both the spatial and spectral resolutions have been enhanced, in comparison to the original images. Dual-Tree Complex Wavelet Transform has the following properties: approximate shift invariance, good directional selectivity in 2-D with Gabor-like filter also true for higher dimensionality: perfect reconstruction using short linear-phase-filter. Advances in night vision imaging have provided alternative methods since signals from image intensifier and uncooled infrared sensors may be combined with in a single image format, and the spectrum may be efficiently combined. In order to be more convenient for human eyes, the 3-D colour space is presented and some improvement has been done, that is ergonomic colour scheme, which can produce imagery with natural appearance and with colours invariant for change in the environmental conditions. This makes the counter intuitive appearance of scenes more legible.

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