

## SIMILARITY MEASURES OF FULL POLARIMETRIC SAR IMAGES FUSION FOR IMPROVED SAR IMAGE MATCHING

H. Ding<sup>a, b, \*</sup>

<sup>a</sup> School of Geosciences and Info-physics, Central South University, Changsha, China – daisyaza@qq.com

<sup>b</sup> Chinese Academy of Surveying and Mapping, Beijing, China

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### ABSTRACT:

China's first airborne SAR mapping system (CASMSAR) developed by Chinese Academy of Surveying and Mapping can acquire high-resolution and full polarimetric (HH, HV, VH and VV) Synthetic aperture radar (SAR) data. It has the ability to acquire X-band full polarimetric SAR data at a resolution of 0.5m. However, the existence of speckles which is inherent in SAR imagery affects visual interpretation and image processing badly, and challenges the assumption that conjugate points appear similar to each other in matching processing. In addition, researches show that speckles are multiplicative speckles, and most similarity measures of SAR image matching are sensitive to them. Thus, matching outcomes of SAR images acquired by most similarity measures are not reliable and with bad accuracy. Meanwhile, every polarimetric SAR image has different backscattering information of objects from each other and four polarimetric SAR data contain most basic and a large amount of redundancy information to improve matching. Therefore, we introduced logarithmically transformation and a stereo matching similarity measure into airborne full polarimetric SAR imagery. Firstly, in order to transform the multiplicative speckles into additivity ones and weaken speckles' influence on similarity measure, logarithmically transformation have to be taken to all images. Secondly, to prevent performance degradation of similarity measure caused by speckles, measure must be free or insensitive of additivity speckles. Thus, we introduced a stereo matching similarity measure, called Normalized Cross-Correlation (NCC), into full polarimetric SAR image matching. Thirdly, to take advantage of multi-polarimetric data and preserve the best similarity measure value, four measure values calculated between left and right single polarimetric SAR images are fused as final measure value for matching. The method was tested for matching under CASMSAR data. The results showed that the method delivered an effective performance on experimental imagery and can be used for airborne SAR matching applications.

### 1. INTRODUCTION

A major research area in SAR image processing is image matching for 3D reconstruction (Toutin, 2000), etc. Area-based image matching methods can provide dense and accurate matching points. However, they are intolerant to radiometric distortions caused by speckles and imaging geometry (Edwards, 2005). Meanwhile, most SAR image matching methods allow match with single polarimetric or fused multi-polarimetric SAR imagery (Toutin et al., 2010) without making full use of multi-polarimetric SAR imagery information. Therefore, in order to obtain a more accurate and robust matching points of two overlapping SAR images, we take a special interest to the full-polarimetric SAR imagery matching method. In this paper, procedures are described to combine four similarity measures calculated from corresponding logarithmically transformed polarimetric images respectively to ensure that the matching points are determined by all data available. The paper is organized as follows: section 2 introduces the similarity measure for matching, section 3 describes the similarity measures fusion procedure, section 4 shows similarity measures and matching points results from CASMSAR airborne full-polarimetric SAR data, and conclusions are drawn in section 5.

### 2. SIMILARITY MEASURE CALCULATED FROM SINGLE POLARIMETRIC SAR IMAGERY

#### 2.1 Overview

Figure 1 illustrates the single polarimetric SAR imagery matching procedure. The goal of this technique is to achieve similarity measures from overlapping SAR images with same polarimetric mode.

The key function of data pre-processing is to improve the subsequent processes by logarithmically transformation. Then, matching from overlapping SAR images is based on the Normalized Cross-Correlation (NCC) similarity measure. Therefore, the procedure consists of two main parts. Part 1 transforms the origin polarimetric image into logarithmically transformed image, in order to change the multiplicative speckles into additivity ones. Part 2 introduces NCC stereo matching similarity measure into SAR image matching.

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\* Corresponding author

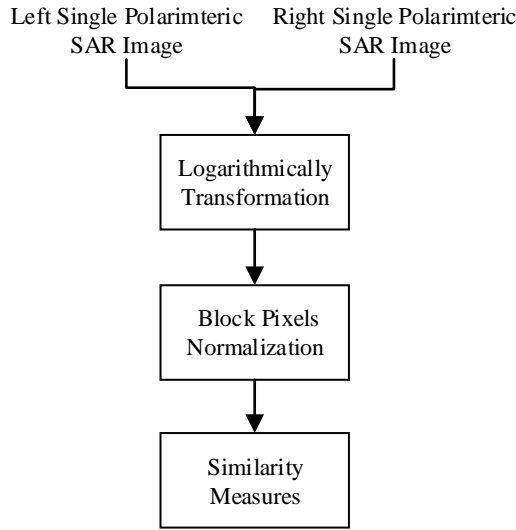


Figure 1. Single polarimetric SAR imagery matching procedure

## 2.2 SAR Imaging Model

Assuming a linear imaging device is used, the intensity  $DN(x, y)$  of a pixel  $P(x, y)$  can be modelled as the function of wave energy backscattered  $P_r(x, y)$  from the corresponding surface (Guo et al., 2000)

$$DN(x, y) = a(x, y)\sqrt{P_r(x, y)} + b \quad (1)$$

where  $a$  = scale factor which includes speckle caused by coherence of radar signal  
 $b$  = additive noise  
 $x, y$  = image coordinates.

As this work is based on the assumption of monostatic radar,  $P_r(x, y)$  is defined by the radar equation

$$P_r = P_t (\sigma^0 A) \left[ \frac{G^2 \lambda^2}{(4\pi)^2 R^4} \right] \quad (2)$$

where  $P_t$  = transmitting power of radar  
 $\sigma^0$  = backscattering coefficient  
 $A$  = scattering cell area  
 $G$  = gain  
 $\lambda$  = wavelength  
 $R$  = slant range between radar and scattering cell area.

The possible additive noise is neglected in this paper because it would complicate the study. We can thus derive  $DN(x, y)$  simply as follows:

$$DN(x, y) = a(x, y)P_r(x, y)^{\frac{1}{2}} \quad (3)$$

Meanwhile, gamma correction is used for compression of the dynamic range which is commonly used before the storage process and its influence shown in equation is as follows:

$$DN(x, y) = a(x, y)P_r(x, y)^{\frac{\gamma}{2}} \quad (4)$$

Where  $\gamma$  = exponent value of gamma correct.

## 2.3 Logarithmically Transformation of SAR Imagery

We assume that the multiplicative speckle in equation (4) is signal-independent. In order to reduce the effects of speckles, logarithmically transformation is employed to transform multiplicative speckles into additive ones (Bovik 2000, Gupta and Gupta 2007). The nonlinear relationship is transformed into a linear one as follows:

$$DN'(x, y) = \log a(x, y) + \frac{\gamma}{2} \log P(x, y) \quad (5)$$

According to equation (2) and (5), the  $\log P(x, y)$  can be specified by the following equation:

$$\begin{aligned} \log P(x, y) &= \log \frac{P_t G^2 \lambda^2}{(4\pi)^2} + \log \frac{\sigma^0(x, y) A(x, y)}{R(x, y)^4} \\ &= \alpha + \beta(x, y) \end{aligned} \quad (6)$$

where

$$\alpha = \log \frac{P_t G^2 \lambda^2}{(4\pi)^2}$$

$$\beta(x, y) = \log \frac{\sigma(x, y)^0 A(x, y)}{R(x, y)^4}$$

The receiving power of radar at image point  $(x, y)$  is determined by many factors including transmitting power of radar, gain, wavelength, backscattering coefficient, scattering cell area and slant range. Thereinto, transmitting power of radar, gain, and wavelength obtained by same SAR sensor are considered to be constant (Guo et al., 2000). However, backscattering coefficient, scattering cell area and slant range are dependent on the incident angle and position of SAR sensor. Meanwhile, different  $(x, y)$  matches different incident angle and position of SAR sensor. Therefore, the  $\beta(x, y)$  values of corresponding pixels in overlapping images obtained from different views are different.

## 2.4 Normalized Cross-Correlation (NCC)

In order to eliminate the effect of transmitting power of radar, gain, and wavelength, we subtract the average of the transformed values in matching window (Block Pixels Normalization), thereby resulting in the removal of  $\alpha$ . It is defined by the following equation:

$$\begin{aligned} DN''(x, y) &= DN'(x, y) - \overline{DN'}(x, y) \\ &= \log a(x, y) + \frac{\gamma(\alpha + \beta(x, y))}{2} - \frac{\sum_{(x, y) \in W} DN'(x, y)}{M \times M} \\ &= \log a(x, y) + \frac{\gamma\beta(x, y)}{2} - \frac{\sum_{(x, y) \in W} \log a(x, y) + \frac{\gamma\beta(x, y)}{2}}{M \times M} \end{aligned} \quad (7)$$

where  $W$  = matching window  
 $M \times M$  = window size

One of the most commonly used SAR image matching similarity measures is Normalized Cross-Correlation (Méric et al, 2011) which is based on an area between two or more SAR images. And it can be stated as follows:

$$NCC(x, y) = \frac{\sum_{j_R \in W_R(x, y)} \sum_{j_M \in W_M(x, y, dx, dy)} DN''(j_R) DN''(j_M)}{\sqrt{\sum_{j_R \in W_R(x, y)} DN''(j_R)^2} \times \sqrt{\sum_{j_M \in W_M(x, y, dx, dy)} DN''(j_M)^2}} \quad (8)$$

Where R = reference image

M = matching image

$dx$  = parallax of range direction

$dy$  = parallax of azimuth direction

The computation of NCC values in each polarimetric band such as HH, HV, VH and VV can be performed similarly. The range of NCC is from -1 to +1.

### 3. SIMILARITY MEASURES FUSION

Under the influence of the geometric distortion, radiometric distortion, and speckles of SAR images, the discriminability of common similarity measures is very low, and most of which are concentrated between 0.2 and 0.3, and NCC similarity measure is no exception. However, SAR images with different polarimetric mode at the same area contain different image information to a certain extent (Zakharov and Toutin, 2009), the accuracy and robustness of matching points can be improved by fusing the similarity measures of different polarimetric images.

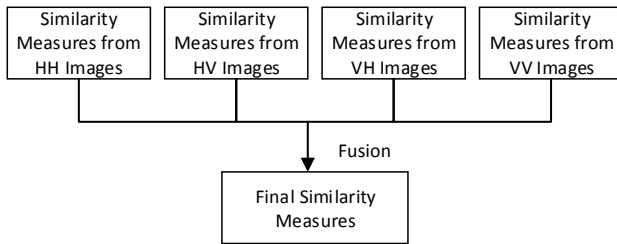


Figure 2. Similarity measures fusion procedure

In this study, similarity measures fusion is performed by information-based entropy technique and the definition of entropy is (Li et al., 2010)

$$Entropy = - \sum_{i=0}^L h(i) \log_2 h(i) \quad (9)$$

Where  $h(i)$  = the probability distribution of intensity  $i$  in the image.

$L$  = the number of pixels in the image.

The relative weight of similarity measure of this polarimetric image is calculated as follows:

$$Scale_i = \frac{E_{Ri} + E_{Mi}}{\sum_{j=1}^4 (E_{Rj} + E_{Mj})} \quad (10)$$

Therefore, the final similarity measure of a point in reference image can be defined as:

$$SM = \sum_{j=1}^4 Scale_i \times NCC_i \quad (11)$$

In general, there must have one maximum correlation coefficient in the curved surface of correlation coefficients and we assume that it shows the right position of the right matching point.

### 4. TEST AND RESULTS

In our experiments, we use airborne X-band full-polarimetric SAR images obtained by CASMSAR system. Point A is illustrated in left reference image with HH polarimetric mode and marked by red cross (see Figure 3). The pixel size in azimuth direction is 0.21m, and in range direction is 0.19m. The flight height is 3415m. The range of matching in right matching image is shown in red rectangle in figure 4. Yellow line in Figure 4 passes through the true matching point in right image, and green line is parallel with the yellow one. They are both edges of objects and there is a remarkable resemblance between them visually, so the similarity measures in these two lines will be close to each other. We fixed the window size to be 15-by-15 pixels.

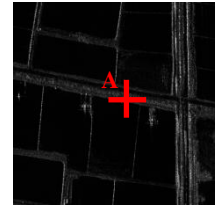


Figure 3. Point A in left image

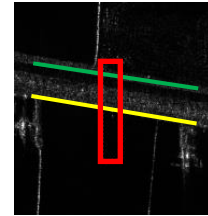
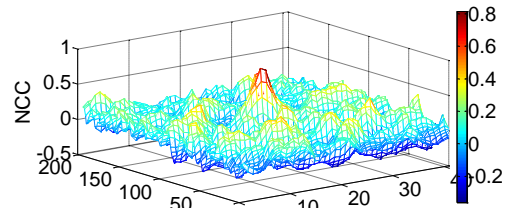
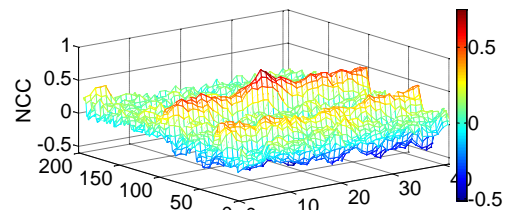


Figure 4. Range of matching in right image

Figure 5 illustrates the curved surface of correlation coefficients between point A in figure 3 and points in rectangle of figure 4 with primary and logarithmically transformed HH polarimetric images respectively. In this search range, the curved surface of correlation coefficients of logarithmically transformed image shown in figure 5(b) is more approximate to real ground based on NCC similarity measure principle especially on the position of green and yellow line. Therefore, logarithmically transformation method weakens the effects of speckles and consequently reveal surface better than primary image.



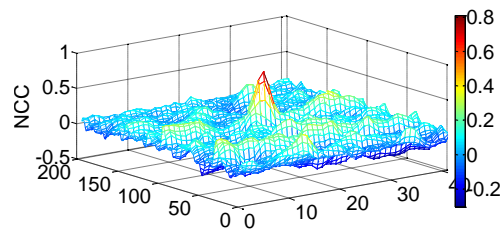
(a) HH polarimetric image + NCC



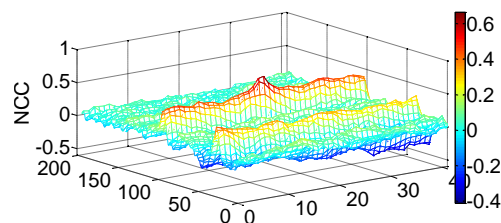
(b) Transformed HH polarimetric image + NCC

Figure 5. Correlation coefficients obtain from primary and logarithmically transformed HH polarimetric images

In order to avoid mismatching and further improve matching accuracy, we take advantage of full-polarimetric data by combining the correlation coefficients of them according to their image entropy. Figure 6 illustrates the curved surface of correlation coefficients between point A in figure 3 and points in rectangle of figure 4 with primary and logarithmically transformed full-polarimetric images respectively. The burr of correlation coefficients is weakened after fusion, and the discrimination is also improved. The maximum correlation coefficient on curved surface shown in figure 6(b) is the correct and precise matching point in right image. The result shows that fusion of similarity measures from full-polarimetric images improve the robustness and confidence of the final similarity measure.



(a) Full polarimetric images + NCC



(b) Transformed full polarimetric image + NCC

Figure 6. Correlation coefficients obtain from primary and logarithmically transformed full polarimetric images

## 5. CONCLUSIONS

The method described in this paper combines similarity measures obtained from four polarimetric SAR images, which makes full use of all available image information. The information is always mutually complementary to a certain extent. Moreover, the effect of speckles is reduced by logarithmically transformation. Therefore, this method leads to a more accurate and robust determination of matching points from overlapping full-polarimetric SAR images than only by single polarimetric images.

Further research is needed to find a more reliable method for similarity measures fusion and increase the discrimination between matching and mismatching points. The method will also be tested on different data and objects.

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