IDENTIFY DAMAGED BUILDINGS FROM HIGH-RESOLUTION SATELLITE IMAGERY IN HAZARD AREA USING DIFFERENTIAL MORPHOLOGICAL PROFILE

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Abstract: This paper presets a methodology and results of evaluating damaged building detection algorithms using an object recognition task based on Differential Morphological Profile (DMP) for Very High Resolution (VHR) remotely sensed images. The proposed approach involves several advanced morphological operators among which an adaptive hit-or-miss transform with varying size, shape and gray level of the structuring elements. IKONOS Satellite panchromatic images consisting of pre and post earthquake site of Sichuan area in China were used. Morphological operation of opening and closing with constructions are applied for segmented images. Unsupervised classification ISODATA algorithm is used for the feature extraction and the results comparison with ground truth data, complex urban area before the earthquake gives 76% and same area wracked after the earthquake gives 88% buildings detection on object based accuracy. This work is being extended to extract shadows and non building objects for better classifications of building roof footprints.

Keywords: Differential Morphological Profile, IKONOS, Building Extraction

1. Introduction

The very high spatial resolution satellite images offer the opportunity to recognize features such as road, vegetation, buildings and other kind of infrastructures. In this paper, we focus on automatic damaged building detection method which is helpful to optimize recognize, rescue, management and recover tasks when an event of hazard. In the past decade, many kinds of methods have developed especially to classification and feature extraction using high-resolution imagery. According to difference manners, these methods can be summarized in to difference kinds: automatic and semi-automatic according to the iteration extent of human; single view and multi views, according to difference principles; region-based and edge-based according to the principle elements acquired manners. Among these methods, mathematical morphology has already proved to be effective for many applications in remote sensing [1] - [9]. Classification and Feature Extraction for Remote Sensing Images From urban Area Based on Morphological Transformations and Classification of Hyper spectral Data From Urban Areas Based on Extended Morphological Profiles were presented by Benediktsson et al [10]. Similarly, Aaron K. Shackelford et al were investigated a method for Automated 2-D Building Footprint Extraction from High-Resolution Satellite Multispectral Imagery [11]. This research also focused on region based classifications.

2. Objectives and Methodology

2.1 Differential Morphological Profile

The Here Differential Morphological Profile (DMP) was developed by feature detectors attempt to identify buildings, shadows, roads and so on of the high-resolution panchromatic images and it is constructed using morphological opening and closing by reconstruction operators. Mathematical morphology employs a set of image operators to extract and analyze image components based on shape and size of quasi-homogeneous regions in the image. This concept is used to create a feature vector from a single image, *I* and it is based on the repeat use of the opening and closing operators, which are commonly used in mathematical morphology [12]. Opening and closing by reconstruction is obtained following by erosion and dilation under the original image [10]. The gray-scale reconstruction $\rho^{*f(p)}$ of image *I* could be defined as follows. Opening γ is defined as the result of erosion followed by the dilation.

$$\gamma^*_{\mathrm{N}}\ f(p) = \rho^{*f(p)}\ (\varepsilon_N f(p)) = \mathrm{Re}\,c(\varepsilon_N f, f) \qquad (1)$$

In a similar fashion, closing f by reconstruction can be defined as

$$\varphi^*_{\mathrm{N}}\ f(p) = \rho^{*f(p)}\ (\delta_N f(p)) = \mathrm{Re}\,c(\delta_N f, f) \qquad (2)$$

Here in the Euclidean transforms assume that flat structuring element that corresponds to the neighborhood SE= $N_G(p)$.The erosion ε_N of the grey level function using the structuring element *N* is defined by the infimum of the values of the grey level function in the neighborhood

$$\varepsilon_N f(p) = \{\wedge f(p^/) \big| p^/) \in N_G(p) \cup f(p)\} \qquad (3)$$

And the Dilation δ_N is similarly defined by the supremum of the neighboring values and the value of as

$$\delta_N f(p) = \{\vee f(p^/) \big| p^/) \in N_G(p) \cup f(p)\} \qquad (4)$$

Opening and closing by reconstruction can be considered as lower-leveling opening and upper-leveling closing operations [13]. The idea of the multi-scale segmentation based on the derivative of the morphological profile was developed as

Let γ*λ be a morphological *opening operator by reconstruction* using structuring element SE = λ and Πγ(x) be the *opening profile* at the pixel x of the image *I* . Πγ(x) is defined as a vector

$$\Pi\gamma(x) = \{\ \Pi\gamma\lambda : \Pi\gamma\lambda = \gamma^*\lambda\ (x),\ \forall\lambda \vee [0,n]\} \qquad (5)$$

Also, let f*λ be a morphological *closing operator by reconstruction* using structuring element SE = λ . Then, the *closing profile* Πfλ at pixel x of the image is defined as the vector

$$\Pi f(x) = \{\ \Pi f\lambda : \Pi f\lambda = f^*\lambda\ (x),\ \forall\lambda \vee [0,n]\} \qquad (6)$$

In the above, Πγ0(x) = Πf0(x) = *I*(x) for λ=0 by the by the definition of opening and closing by reconstruction [3]. Given (1) and (2), the opening profile can also be defined as a granulometry [1] made with opening by reconstruction, while the closing profile can be defined as antigranulometry made with closing by dual reconstruction. The derivative of the morphological profile is defined as a vector where the measure of the slope of the opening-closing profile is stored for every step of an increasing SE series. The *derivative of the opening profile* Δγ(x) is defined as the vector

$$\Delta\gamma(x)\ = \{\ \Delta\gamma\lambda\ :\ \Delta\gamma\lambda\ =\ |\Pi\gamma\lambda - \Pi\gamma\lambda\text{-}1\ |,\ \ \forall\lambda \vee [1,n]\} \qquad (7)$$

By duality, the *derivative of the closing profile* Δf(x) is the vector

$$\Delta f(x) = \{\ \Delta f\lambda : \Delta f\lambda = |\Pi f\lambda - \Pi f\lambda\text{-}1\ |,\ \forall\lambda \vee [1,n]\} \qquad (8)$$

Generally, the *derivative of the morphological profile* $\Delta(x)$or the DMP can be written as the vector

$$\Delta(x) = \begin{array}{l} \Delta c : \Delta c = \Delta f \;\; \lambda = n-c+1, \; \forall c \vee \;\; [1,n] \\ \Delta c : \Delta c = \Delta\gamma\lambda = c-n, \; \forall c \vee \;\; [n+1,2n] \end{array} \qquad \{ \; (9)$$

with equal to the total number of iterations, c=1,…..,2n, and , and |n-c|=the size of the morphological transform.

2.2. Used Data

The morphological filter theory was designed for a series of gray-level images. In this paper, we used pre and past IKONOS panchromatic (PAN) imagery of Sichuan earthquake in China in 2008. The gray color image of high-resolution IKONOS panchromatic images that consist of 1m resolution band (450-900 nm) is used. The images were smoothed using median convolutions filter for removal post classification procedures like “salt-and-paper” and other visual enhancement procedures. IDL programming language and ENVI 4.7 commercial software package is used for image processing and classification on this research.

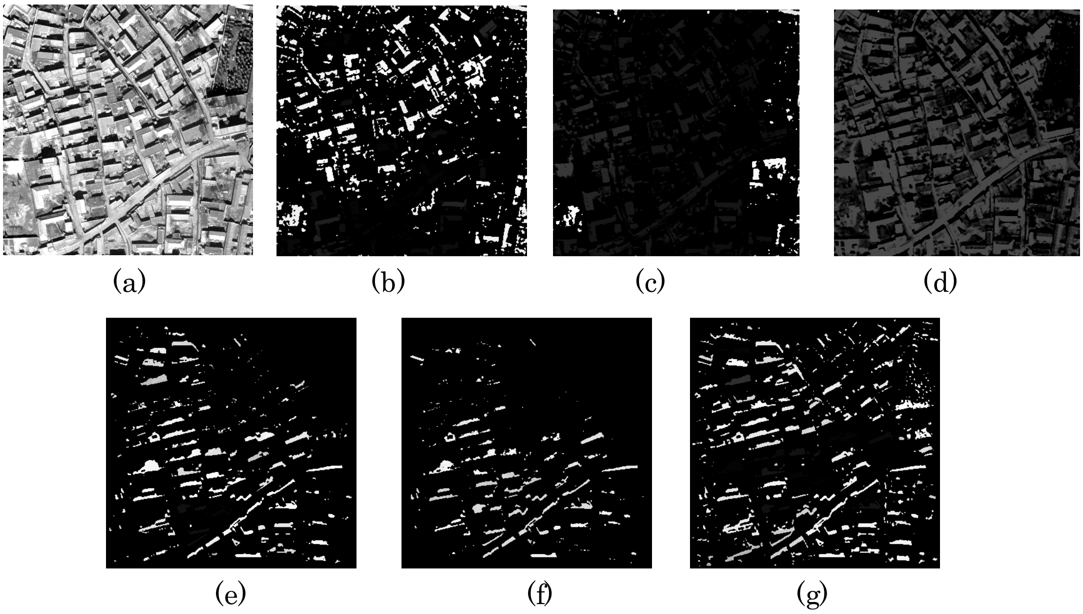


Figure 1. *The image (a) shows the original image before the earthquake event and (b)-(g) represents Structural decomposition of the image using differential morphological profile. The images have been visually enhanced. The derivative has been calculated relative to a series generated by six iterations of the elementary SE with radius from 7-19m. Derivative of the opening profile with r=(b)7, (c)11, (d)15 and closing profile with r=(e)7,(f)11,(g)15 are shows above respectively.*

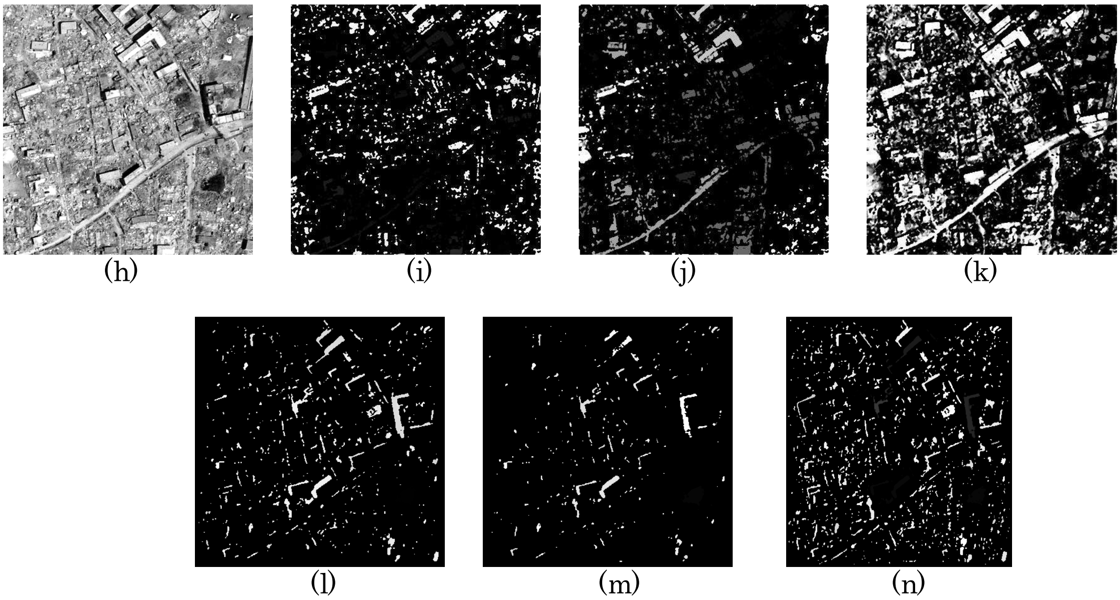


Figure 2. *The image (h) shows the original image after the earthquake event and (h)-(n) represents Structural decomposition of the image using differential morphological profile. The images have been visually enhanced. The derivative has been calculated relative to a series generated by six iterations of the elementary SE with r (radius) from 7-19m. Derivative of the opening profile with r=(i)7, (j)11, (k)15 and closing profile with r=(l)7,(m)11,(n)15 are shows above respectively.*

3. Building Extraction

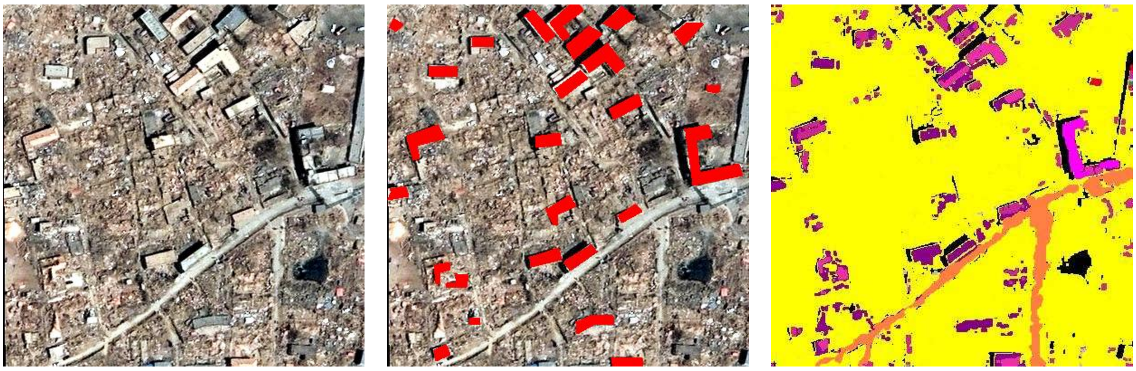
The building shadows are easy to extraction using their low reflection value. Musk for the shadows is built up using the reflection value between 0 and 60. The structures with similar scale to the SE diameter give high response when SE in DMP value with bright structures in opening portion and dark structures in closing portion of the profile. For each pixel in the image, the position of the maximum response within the DMP vector ($\Delta(x)$), indicate both the SE size that best characterize the structure that the pixel resides within and whether the pixel is part a structure that is brighter or darker than the surrounding region. The maximum DMP response indicates with well match SE value that the pixel resides within. There are 8 differential morphological profiles were created using disc shaped morphological elements with radius (r) increasing 7 to 19m (step size is equal to 4m). The SE that less than 7m are not reliable for use because of it consists of small shadows, trees and wracked of buildings. Those fingers give noise for the classification results, we used that SE more than 7 to detect for remain Buildings. Most of the bright building roof is gives the maximum response with opening differential profile and dark color roof, shadows are with closing differential profile. Unsupervised ISODATA Classification way is used for the classification and identification the structures. Combining morphological operations is carried out for remove pixel errors that occurred due to delineation of image objects with DMP before classification.

4. Results

The results have shown the usefulness of the proposed method during detection of various types of building, as illustrated by the portions given this paper. The image patches used to train the Unsupervised ISODATA classifier. The shadows of the building were masked when classification using their low spectral value. The candidate area contained various kinds of roofs with difference colors and shapes before the earthquake. There are some building structures that complex and combine together were classified as a one building. Although this result appeared to be high accuracy, the confidence measures produced by the ISODATA training suggested a reliability of post event gives 88.46%. The error extraction of building structure could be due to over fitting of the decision surface to the data. The totally collapsed building in the applied area is identified as 89 according to manually labeled buildings as ground truth and the result of the above algorithm views as 65.

(o) (p) (q)

Figure 3. *Building extraction results. (o) IKONOS image of the pre-earthquake area. (p) Manually labeled buildings as ground truth. (q) Result of the building extraction according to approached method.*



(r) (s) (t)

Figure 4. *Building extraction results. (r) IKONOS image of the post-earthquake area. (s) Manually labeled buildings as ground truth. (t) Result of the building extraction according to approached method.*

Table1. *Before the earthquake*

	Object base	Pixel base
Correctly Extracted Buildings	88	62572 (39.11%)
Total	115	79240 (49.56%)
	76.52%	Dismiss pixels 16668

Table2. *After the earthquake*

	Object base	*Pixel base*
Correctly Extracted Buildings	*23*	*12886 (08.05%)*
Total	*26*	*7308 (04.56%)*
	88.46%	*Overlap pixels 5578*

5. Conclusions

We applied a method for extraction of urban structures and hazard estimation using very high-resolution satellite images. The first step was to segmentation structural information using morphological opening and closing by reconstruction operators. IKONOS satellite panchromatic gray level images of pre and post earthquake event were applied to morphological operators. Then, the building footprint were extracted in candidate region using connected components analysis to the pixels selected according to their morphological profiles, obtained using increasing structural element sizes for 7 to 19m for opening and closing operators. From the analysis of the extraction results, complex urban area before the earthquake gives 76% and same area wracked after the earthquake gives 88% buildings detection on object based accuracy percentage according to the applied method. Further work is required to increase the accuracy of building detection and determine if damage ratio of the structure can be estimated.

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Acknowledgements

We would like to thank for access and imagery provided by GeoEye (space imagine) and Global Land Cover Facility (GLCF).

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