

Road Map Extraction from Satellite Images Using Connected Component Approach

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Abstract: The satellite images usually consist of complexity due to presence of several manmade and natural features. Automatic extraction of features from satellite images is a significant analysis for getting an efficient geographical database system and modifying the remote sensing and geographic information system (GIS) areas. We have presented a methodology using image processing and connected component approach for automatic extraction of road network from any high resolution satellite image. Its quite challenging in the areas having complex scenes. There is an algorithm to extract the manmade features from the image and filtering operations are used to correctly extract out the respective feature i.e. the road network from the image. The proposed method follows various steps consisting of Enhancement, Filtering operations, Segmentation etc. The extracted road network can be used for the map creation which shortens the human efforts in creating large scale maps and also it is very efficient in time and cost. The method is simplistic to implement in real time also and can be applied for the real time development in the field of remote sensing and Geographic Information System including rural and urban development.

Keywords: Connected component, Enhancement, Feature extraction, Filtering operations, Road network detection.

I. INTRODUCTION

Now-a-days automatic detection of Road networks from satellite images is a most prominent research subject and it is widely used for various Remote sensing applications. It basically generates geospatial information for large areas with very minor inaccuracy. Moreover cost and time of data production using these methods are far lower than almost all other production methods. Roads recognition is unavoidable as they form an important Geographic Information System (GIS) layer in remarkable civilian and military applications involving navigation or Location aware systems and emergency planning systems for evacuation and Fire response [1].

Basically improvement of pictorial information for human interpretation is one of the major application areas in image processing. A number of satellite original images contain noise. This can be eliminated by several enhancement techniques. Filtering is an enhancement technique, used to remove or filter unnecessary information (noise) from the image. It is also applied for image smoothing and sharpening. Few neighbourhood operations work with the values of the image pixels in the neighbourhood and the respective values of a sub image having the same dimensions. Picture enhancement and its gray level adjustment are the primary steps towards the fulfilment of the whole process. After converting the image into gray scale image its effects can be analysed using histogram analysis. Using histogram, we can analyse the black and white proportions of the image. Once image is converted into gray scale form, its level is adjusted in such a manner that black and white proportions of the image could be separated out up to a certain level so that thresholding in the next step might become easier as our primary task is to provide approximate binary image

before reaching to thresholding. GIS needs the automatic extraction process of roads for their data updating because it is very costly and time consuming process to update data to their database manually also there may be a possibility of error in manual updating of road network. There are variety of services facilitated by GIS applications containing road databases include satellite navigation, creation and updating of accurate geographical databases,, route planning, health-care accessibility planning, military operations, infrastructure management, land cover classification and also the extent of damages after natural disasters such as floods, earthquakes [2]. The aim of road network extraction is just to provide a binary form which consists of two regions i.e. true region representing the road region and the false region representing non-road regions but the problem arises when there are complex structures in the image containing several different objects, such as roads, trees, vehicles, houses, etc., with differences in their shapes and textures. An image contains number of connected components (CC), and each CC is a set of pixels where each pair of pixels is connected to one another. There is also a method called Trivial opening which is used here to filter these sets of CC of the image based on particular conditions. If this specific condition is true then connected component is retained otherwise, it is removed. We may use trivial opening only when the concerned image is having the connected components [3,4]. Proposed method can be implemented on noisy images too, and good result can be obtained. In this paper connected component and thresholding based algorithm is applied for road network extraction, Second stage includes methodology and system algorithm, third stage relates with connected component approach. Fourth and fifth

stages are concerning with experimental results, performance measures and conclusions parts respectively.

II. SYSTEM MODEL

The system model of the road network extraction consists of various steps. In this a satellite image is loaded first, then various enhancement methods are performed to improve the quality of the image before proceeding to the next steps. After that image is converted into gray scale image processing. Subsequently other operations are performed as the figure 1 as per the complexity and noise level of input image.

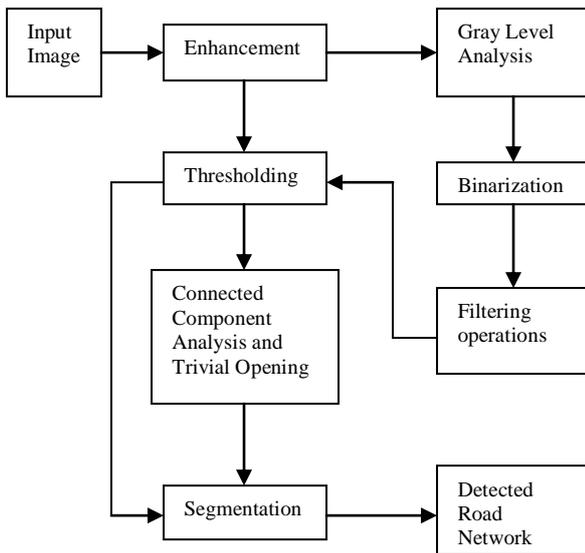


Fig1. Road Network Extraction Algorithm

The figure 1 illustrates the road network extraction algorithm which includes various steps which are performed over the input test satellite image. The details of each step are presented below

(a) The input image is loaded. It is called the test image and its pixel intensity is analysed by obtaining its histogram, shown in figure 2 (a & b)

Now Image is enhanced by applying the enhancement step and its aim is to improve the visual effects in the image. Thus it helps geographic image interpretation and improves the contrast and brightness between the target (road regions) and non-target (non-road regions) for high level image processing. Enhancement operation causes in the internal regions of the manmade structures being more homogeneous than the border regions of the image [1].

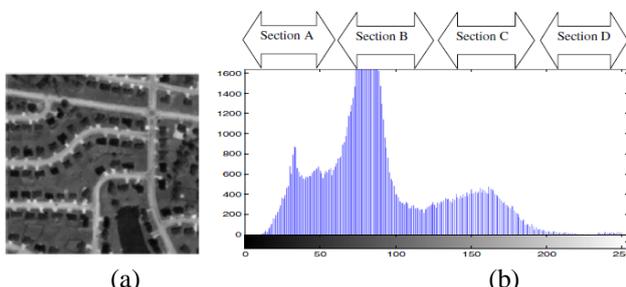


Fig. 2. (a) Input Image. (b) Image Histogram

(b) Now after the application of enhancement method the manmade objects appear more prominent in the image than the natural objects. Hence it facilitates now to apply gray scale conversion on the enhanced image because it contains fewer natural objects as per our need and image is converted to gray scale form. [1]

(c) The next step is binarization in which gray scale image is converted to binary image or black and white image and further it yields to thresholding. Adaptive global thresholding is applied here to remove the non-road pixels and segment approximated road region from the satellite image.[5]. Meanwhile if image contains noise then filtering operation can be performed to remove the noise from the image.

Thresholding is applied on the binary image. A suitable threshold value is selected to segment the approximated road regions from the image. Otsu's algorithm is applied to find the threshold value V_T which is useful to segment the road region pixels $r(x,y)$ from the original image pixels $f(x,y)$.

$$r(x,y) = \begin{cases} 0, & \text{if } f(x,y) < V_T \\ 1, & \text{if } f(x,y) \geq V_T \end{cases}$$

The histogram shown in the figure 2 (b) of the satellite image is analyzed and categorised into four principle sections to obtain the desired threshold value for the segmentation, The histogram is divided into sections based on the average value (M) of all the pixel intensity values in the image. Section A shows the pixels having the intensity values lies between the lowest intensity values to half of the average value (M). This section recognizes dark objects such as vehicles, lakes ,shadows etc in the image. Section B relates with the pixels having intensities from half of the M value to M value. This section provides the information of dark gray shade objects like grasslands, trees, etc. Section C finds the pixel values from M to half of maximum intensities, it belongs to bright gray objects for instance roads, lanes, etc. Section D relates with the group of pixels having the values between half of the maximum intensity to maximum intensity and this section finds the bright objects such as bright vehicles, concrete cement road, etc. Adaptive thresholding technique is used in this case because single value thresholding might not work here properly as threshold value of pixel is based on its position within an image. From this technique we can extract approximate road regions [5].

III.CONNECTED COMPONENT APPROACH

Connected component of an image may be defined for any pixel in that image, as a set of pixels that is connected to that pixel [6].The set of pixels, not separated by a definite boundary can be featured as connected pixels. Connected component is actually each maximal region of connected pixels . An image is divided into segments by connected components. Connected component approach is applied for feature extraction, edge detection, line detection, road map extraction, etc. We can extract connected components in an image using dilation and intersection operators..

Suppose Z be an image and connected component in an image is denoted as A; and a known point of A is denoted as p which is referred as X0. Then, the expression given below gives all elements of A.

$$X_k = (X_{k-1} \oplus B) \cap Z \quad (1)$$

Here k belongs to positive integers, symbol \oplus represents morphological dilation, \cap denotes intersection and B is a suitable structuring element. If $X_k = X_{k-1}$, then the algorithm has converged and let $A = X_k$.

The connected components that are based on some criterion are extracted by trivial opening [7]. Suppose Z be an image $\{Z(n) | n=1,2,3,\dots,N\}$ is a sequence of connected components in the image Z, and z(i) is a point in Z(i). The trivial opening can be expressed with a condition S, as follows in equation 2.

$$s_o(z) = \begin{cases} Z(i), & \text{if } Z(i) \text{ satisfies the condition } S \\ \emptyset, & \text{otherwise} \end{cases} \quad (2)$$

Here, $s_o(z)$ is the trivial opening with condition S associated. Trivial opening is the method, which is useful to extract the required connected components from the image based on condition S and remove the connected components which do not satisfy the condition S. Road regions can be easily identified and extracted using trivial opening. We can extract long feature roads by selecting the condition as the long axes of minimum ellipse for trivial opening mechanism. Using equation given below, we can obtain trivial opening for road detection [7].

$$R_o = \left\{ P \mid \begin{array}{l} \text{Long axis of minimum ellipse} \\ \text{enclosing } P(i) \geq S \end{array} \right\} \quad (3)$$

Where P is an image and P(i) is a connected component of the image. We may proceed by extracting all the connected components from the image by using equation 1. Out of these connected components, further road components are segmented using the equation 3. Thus, the resultant image R_o after the analysis has the required connected road component which is greater than S.

IV. EXPERIMENTAL RESULTS AND DISCUSSION

The proposed method is applied on test image which is a satellite image, collected from Satellite Imaging Corporation and Apollo Mapping [8, 9]. The experiment is performed on the same image and result is obtained after the segmentation using the proposed method and the road network is detected. Corresponding test image and results obtained after implementing adaptive thresholding, connected component analysis and trivial opening are shown in the figure 3. Final road map can be extracted using the morphological thinning operation on the resulting image.

Performance of the experiment is analyzed using basically three parameters which are very useful in road network extraction which are known as Completeness, Correctness and Quality.

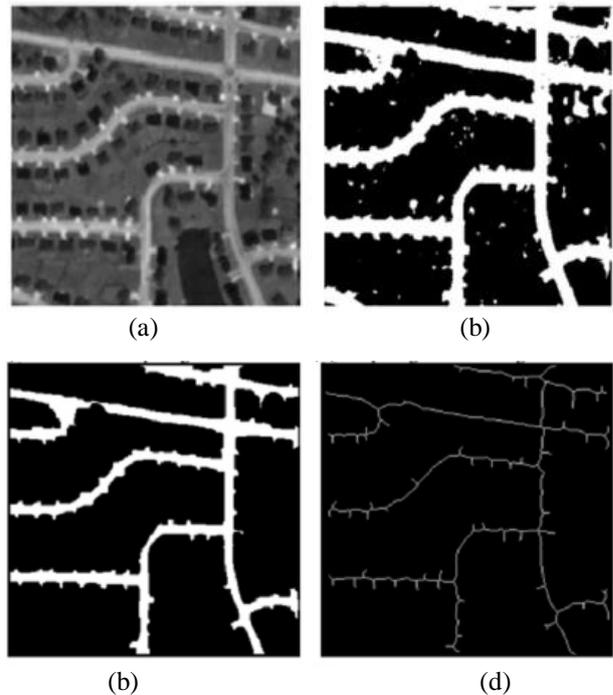


Fig. 3. (a) Input test satellite image (b) Result after thresholding (c) Road regions segmented after trivial opening (d) Extracted Road Network.

These parameters are measured using a ground truth road map which is manually drawn and the extracted road network is compared with it. We form a buffer around this reference map with constant width.

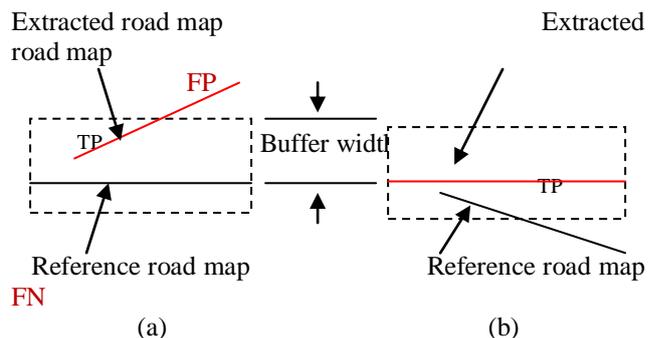


Fig. 4. (a) Matched extraction (b) Matched reference

According to figure 4, the matched road data in extracted map is called true positive (TP) and unmatched road data is False positive (FP). The reference data within buffer is called matched reference. The length of matched portion of reference with extracted road data is TP while unmatched reference is called False Negative (FN) data.

$$Compl \% = \frac{\text{length of the matched reference}}{\text{length of reference}}$$

$$\cong \frac{TP}{TP + FN} \times 100$$

$$Correct \% = \frac{\text{length of the matched extraction}}{\text{length of extraction}}$$

$$\cong \frac{TP}{TP + FP} \times 100$$

$$\text{Qual \%} = \frac{\text{length of matched extraction}}{\text{length of the extracted data} + \text{length of unmatched reference}}$$

$$\cong \frac{TP}{TP + FP + FN} \times 100$$

Using above equations average measures of the Completeness, Correctness and Quality are determined which are 89%, 94% and 87% respectively. For evaluation, true positive, false positive and false negative values can be calculated and counted using manually drawn ground truth map which is obtained using the software GIMP (GNU Image Manipulation Program) [10]. Buffer width is set as 3. Value of TP, FP and FN is counted on the ground truth map as follows.

TP- It is the element which is present on ground truth and the extracted road network both.

FP- It is present on the extracted road network while not on the ground truth.

FN- It is the element present on ground truth while not on the extracted road network.

Table I. Comparison of average measured values of proposed method with other methods

Methods	Compl %	Correct %	Qual %
Xiaoying [11]	78	80	66
Wenzhong [12]	79	77	63
Proposed work	89	94	87

Table I shows average measured values of completeness, correctness and quality of the proposed method are compared with other methods and it can be analyzed that proposed method performs very well over these methods and good results are obtained using proposed method.

V. CONCLUSION

We have proposed a method which is applied for road network extraction from satellite images. Road network extraction is very important in remote sensing and GIS applications where information keep on updating regularly. Using connected component approach and thresholding we obtain good results in terms of correctness, completeness and quality. We can apply proposed method on noisy images also.

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REFERENCES

- [1] D. Chaudhuri, N. K. Kushwaha, and A. Samal, "Semi-automated road detection from high-resolution satellite images by directional morphological enhancement and segmentation techniques," in Proc. IEEE J. Sel. Topics Appl. Earth Observ. Remote Sens., vol. 5, no. 5, pp. 1538–1544, Oct. 2012
- [2] N. Lomenie, J. Barbeau, and R. Trias-Sanz, "Integrating texture and geometric information for an automatic bridge detection systems," in Proc. IEEE Int. Geosci. Remote Sens. Symp. (IGARSS), Toulouse, France, Jul. 2003, pp. 3952–3954.
- [3] MD Mura, JA Benediktsson, B Waske, L Bruzzone, Morphological attribute profiles for the analysis of very high resolution images. IEEE Trans. Geosci. Remote Sens. 48(10), 3747–3762 (2010)
- [4] M Boldt, K Schulza, A Thiele, S Hinz, Using morphological differential attribute profiles for change categorization in high resolution SAR images. Inter. Arch. Photogramm. Remote Sens. Spatial Inf. Sci. 1(1), 29–34 (2013)
- [5] PP Singh, RD Garg, Automatic road extraction from high resolution satellite image using adaptive global thresholding and morphological operations. J. Indian Soc. Remote Sens. 41(3), 631–640 (2013)
- [6] MD Mura, JA Benediktsson, B Waske, L Bruzzone, Morphological attribute profiles for the analysis of very high resolution images. IEEE Trans. Geosci. Remote Sens. 48(10), 3747–3762 (2010)
- [7] J Serra, L Vincent, An overview of morphological filtering. Circuits, Syst. Signal Proc. 11(1), 47–108 (1992)
- [8] Satellite images from Apollo Mapping. Available: <https://www.apollomapping.com>
- [9] Satellite images from Satellite Imaging Corporation (SIC). Available: <http://www.satimagingcorp.com>
- [10] RW Solomon, Free and open source software for the manipulation of digital images. Am. J. Roentgenol. 192(6), 330–334 (2009)
- [11] X Jin, CH Davis, An integrated system for automatic road mapping from high-resolution multi-spectral satellite imagery by information fusion. Inf.Fusion 6(4), 257–273 (2005)
- [12] W Shi, Z Miao, Q Wang, H Zhang, Spectral-spatial classification and shape features for urban road centerline extraction. IEEE Trans. Geosci. Remote Sens. Lett. 11(4), 788–792 (2014)