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# 2 **Building Extraction from very high resolution stereo** 3 **satellite images using OBIA and topographic** 4 **Information**

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12 **Abstract:** The availability of very high resolution (VHR) satellite imagery (<1m) has opened new  
13 vistas in large scale mapping and information management in urban environments. Buildings are  
14 the most essential dynamic incremental factor in urban environment and hence their extraction is  
15 the most challenging activity. Extracting the urban features particularly buildings using traditional  
16 pixel based classification approaches as a function of spectral tonal value produces relatively less  
17 accurate results for these VHR Imageries. The present study demonstrates the building extraction  
18 using Pleiades panchromatic (PAN) and multispectral stereo satellite datasets of highly planned  
19 and dense urban areas in parts of Chandigarh, India. The stereo datasets were processed in  
20 photogrammetric environment to obtain the Digital elevation Model (DEM) and corresponding  
21 orthoimages. DEM's were generated at 0.5 m and 2.0 m from stereo PAN and multispectral datasets,  
22 respectively. The orthoimages thus generated were segmented using Object Based Image Analysis  
23 (OBIA) tools. The object primitives such as scale parameter, shape, textural parameters, and DEM  
24 derivatives were used for segmentation and subsequently to determine threshold values for  
25 building fuzzy rules for building extraction and classification. The rule based classification was  
26 carried out with defined decision rules based on object primitives and fuzzy rules. Two different  
27 methods were utilized for performance evaluation of the proposed automatic building approach.  
28 Overall accuracy, correctness and completeness were evaluated for extracted buildings. It was  
29 observed that overall accuracy was higher (> 93%) in areas having larger building and sparse built-  
30 up as compared to areas having smaller buildings and dense built-up.

31  
32 **Keywords:** Object based Image Analysis; High Resolution Image processing; Building Extraction;  
33 Textures.  
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## 35 **1. Introduction**

36 Rapid growth in the urbanization, demands monitoring features and a decision making system  
37 for planning and utility mapping. Feature extraction is an important aspect of remote sensing and is  
38 becoming more important than ever due to large number of datasets coming each day, which require  
39 extensive processing and automation. Subsequently, manual extraction and interpretation is time  
40 consuming. Automatic feature extraction has become a main objective to save time in  
41 updating data, and as a result of this, people are changing progressively to new

42 procedures [1], specially focusing on extracting Roads [2] and Buildings [3,4]. Extraction of buildings  
43 on a common platform using high resolution (HR) satellite imagery is a complex & challenging task  
44 due to high variability in building features. A wide range of techniques and algorithms have been  
45 proposed for automatically constructing 2D or 3D building models from aerial imagery [5,6].

46 Quantitative information about building density can serve as a useful tool for the study of land  
47 use changes, illegal building development, urban expansion, etc., and, therefore, can contribute to  
48 the sound development of an area. Typical urban structures often have sizes smaller than the spatial  
49 resolution of the sensor when they are represented on a medium resolution image. In a high spatial  
50 resolution different texture values characterize them and can be used for identifying building and  
51 their densities. Rational polynomial coefficients (RPCs) have been a successful alternative to physical  
52 sensor model in relating the image space with the ground space [7, 8]. DEMs are an important  
53 requirement for the generation of high quality orthoimages and preparation of maps [9-12]. HR  
54 images whose spatial resolutions are about 1m were assessed for performance of the generated  
55 rational function model (RFM ) generation process as well as the physical sensor model. The stereo  
56 pair's horizontal positioning accuracy of the physical and RPC models was found as 5.0 m in both  
57 cases with circular error 90 percent (CE90) [13].

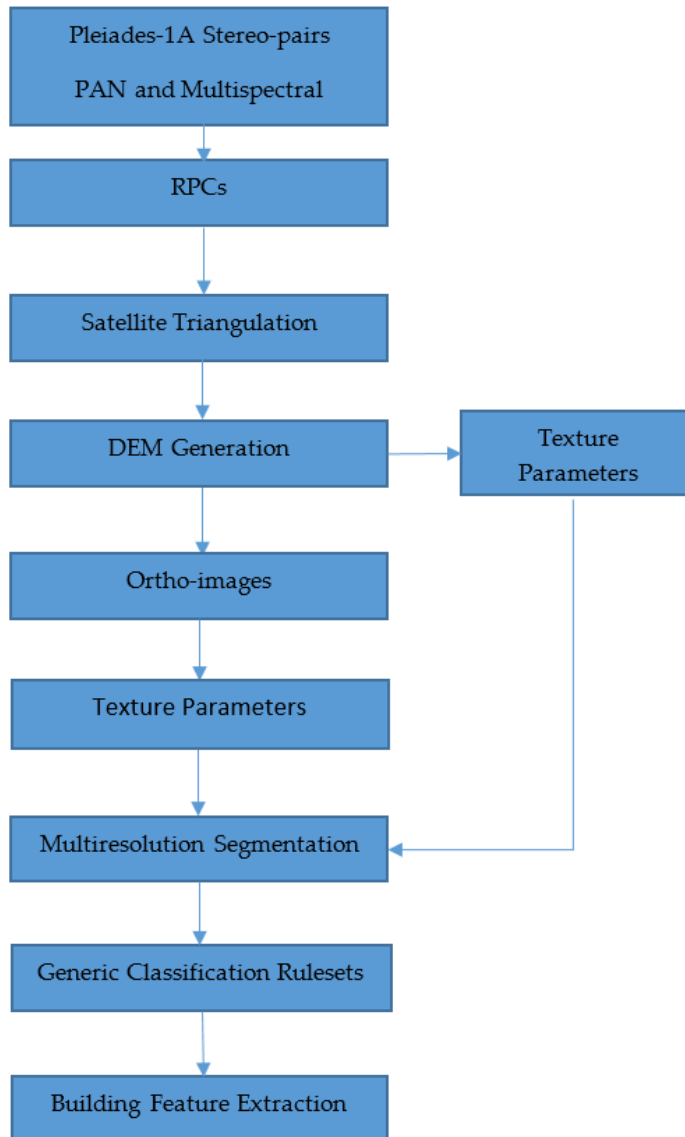
58 Detecting building objects is a challenging topic even with high-quality remote sensing data.  
59 Rooftops have many forms and appearances regarding their color, shape, material, and optical issues  
60 such as illumination, reflection and perspective also bias the automatic identification. Although many  
61 studies have focused on feature extraction, they use spatial or structural features or both [14]. There  
62 are very few studies on incorporation of DEM in building extraction. An attempt is made here to  
63 utilize DEM, its derivatives, textures along with spectral and spatial information in an object based  
64 environment for building extraction.  
65

## 66 2. Materials and Methods

67 VHR and HR satellite Pleiades-1A panchromatic (PAN) and multispectral stereo image datasets  
68 were used in this study. Chandigarh is a city and a union territory of India that serves as the capital  
69 of the Indian states of Punjab and Haryana. Chandigarh is bordered by the state of Punjab to the  
70 north, west and south, and to the state of Haryana to the east. Chandigarh was one of the early  
71 planned cities in the post-independence India and is internationally known for its architecture and  
72 urban design. The master plan of the city was prepared by Swiss-French architect Le Corbusier. It is  
73 located near the foothills of the Shivalik range of the Himalayas in northwest India. The overall  
74 methodology for photogrammetric processing and feature extraction carried out in this study is  
75 depicted in Figure 1.

76 The elevations played a significant role in high spatial resolution image photogrammetric  
77 processing and orthoimage generation The Pleiades stereo data, which was acquired was  
78 photogrammetrically processed for satellite triangulation using the RPCs. Thereafter, the DEM's  
79 were generated for the experimental site using PAN and multispectral stereo pairs at 0.5m and 2m  
80 posting interval respectively. The root mean square error (RMSE) achieved for Panchromatic (PAN)  
81 and Multi-spectral (MS) stereo datasets was 0.05 pixel and 0.03 pixel respectively Further the Digital  
82 elevation Model (DEM) and corresponding orthoimages were generated for both PAN and  
83 Multispectral stereo datasets.

84 The two orthoimages were then fused together using Principal components. An integrated  
85 dataset comprising of resulting Pan sharpened orthoimage, DEM layer, Suitable texture layers from  
86 DEM and orthoimage were then segmented using Multiresolution segmentation technique.. The  
87 object primitives such as scale parameter, shape, textural parameters, DEM derivatives and textures  
88 were used for segmentation and subsequently to determine threshold values for building fuzzy rules  
89 for building extraction and classification. The rule based classification was carried out with defined  
90 decision rules based on object primitives and fuzzy rules.  
91



93 **Figure 1.** Methodology for photogrammetric processing and feature extraction

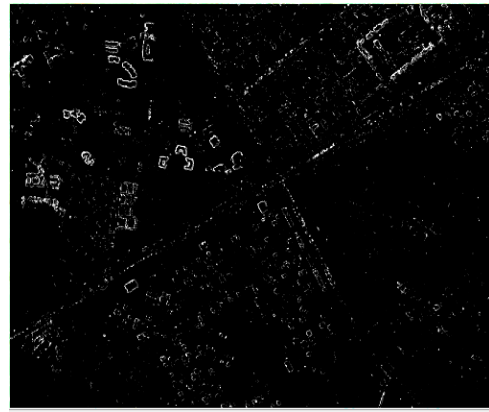
### 94 **3. Results and Discussion**

95 The results are illustrated in Figure 2 and 3. The DEM (Figure 2a) has brighter tones both for  
 96 buildings and Roads. Low elevation areas are in black and agriculture areas, grasslands are grey. It  
 97 was observed that variance, angular second moment (Figure 2c) were able to separate roads from  
 98 buildings; variance, entropy and grey level co-occurrence matrix (GLCM) correlation (Figure 2d) were  
 99 useful in extracting building footprint. The PAN sharpened image is presented in Figure 3a and  
 100 variance of the same in Figure 3d also indicates that buildings and roads are distinguishable from  
 101 other features using this texture. The multiresolution segmentation using fractal net evolution  
 102 approach (FNEA), scale parameter converged to 220 which was found suitable for both larger and  
 103 smaller buildings. Since the main aim was to extract urban features mainly buildings, vegetation  
 104 (Trees and grass land) was eliminated using normalized difference vegetation index (NDVI). The  
 105 quality measures were computed which address the problem of completeness and correctness of the  
 106 extracted road network. The results attained are completeness 93.47%, correctness of 92.94% with the  
 107 overall quality being 90.79%.

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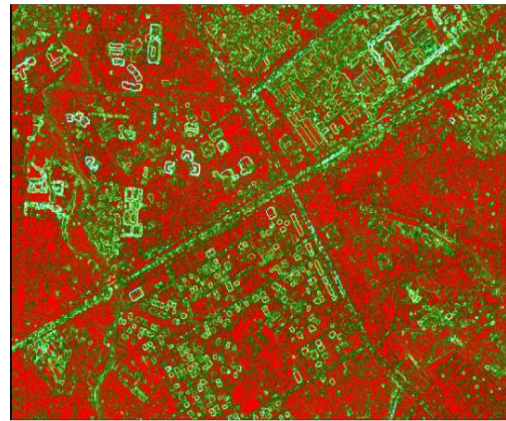
(a)



(b)



(c)



(d)

109

110

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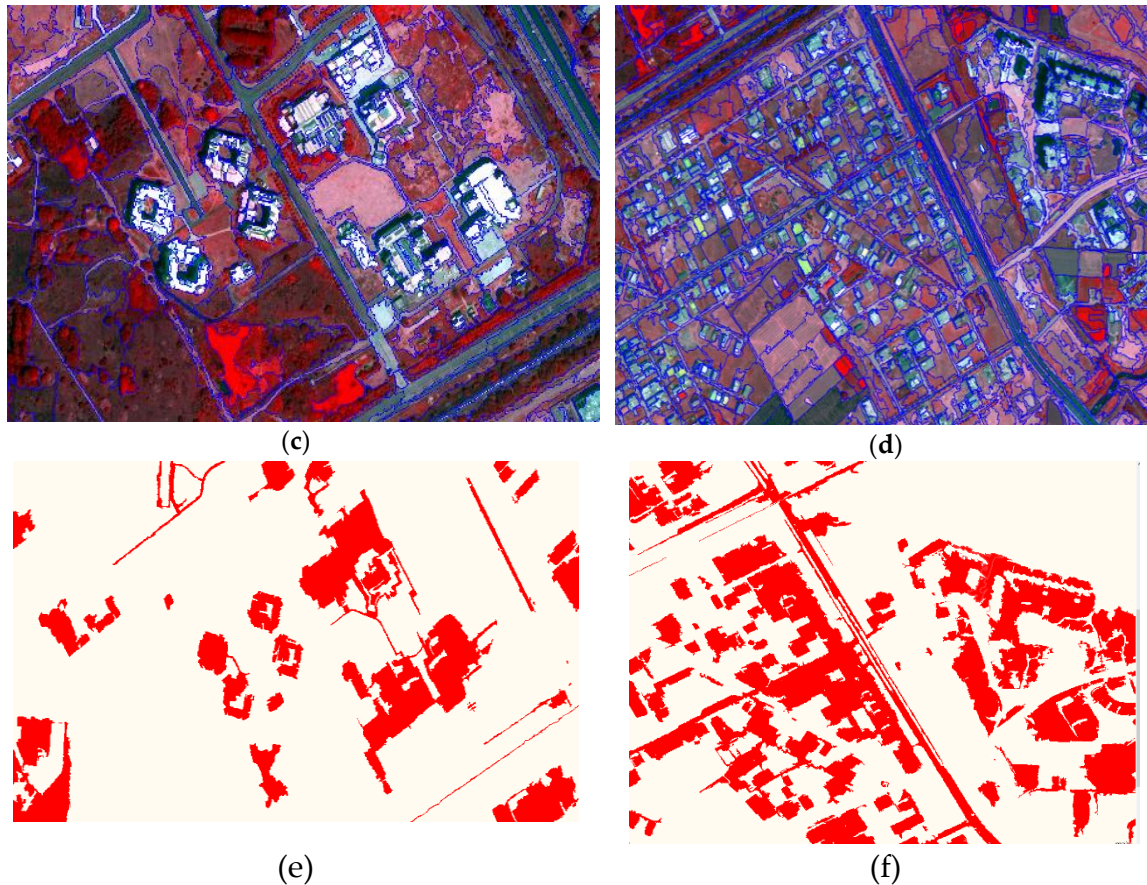
**Figure 2.** (a) DEM, (b) Variance, (c) Angular Second Moment and (d) FCC (R: GLCM Correlation, G: Entropy B: Variance)



(a)



(b)



112 **Figure 3:** (a) Fused FCC Orthoimage, (b) FCC-GLCM Variance, (c, d) Segmented Images (e, f)  
 113 Extracted Buildings

114 **4. Conclusions**

115 As evident from the achieved RMSE values for the satellite triangulation, the bundle block  
 116 adjustment based on RFM is of good quality. The block is thus able to produce good DEM and  
 117 orthoimages for the feature extraction as observed from visual interpretation of clearly identifiable  
 118 drainage networks and built-up regions in the DEM and orthoimages. Generic Rulesets which can be  
 119 transferable for each and every dataset can have the limitations of terrain areas. Therefore, the  
 120 rulesets are having restriction in different feature extraction. Slope, Altitude and nature of terrain  
 121 plays important role in urban feature Interpretation. In this paper, an approach to extract buildings  
 122 road from very high resolution orthoimage using was carried out using a DEM and various texture  
 123 parameters extracted from DEM and raw images. The method described above worked well in those  
 124 area also which contained shadows. The obstacle encountered were shadows of high buildings fall  
 125 on smaller building rooftops and the tone of the road spectrally mixing up to with building rooftops.  
 126 Higher dimension dataset (texture and transforms) were found useful in most of the problems  
 127 encountered. It was observed that overall accuracy was higher (> 93%) in areas having larger building  
 128 and sparse built-up as compared to areas having smaller buildings and dense built-up. The method  
 129 worked well to extract fine building boundaries both in highly dense and rural areas.  
 130

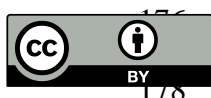
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133 **Author Contributions:** "Both authors conceived, designed performed the experiments, analyzed the data and  
 134 wrote the paper."

135 **Conflicts of Interest:** "The authors declare no conflict of interest."

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