

Proceeding Paper

Land Use and Land Coverage Analysis with Google Earth Engine and Change Detection in the Sonipat District of the Haryana State in India [†]

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[†] Presented at the 9th International Electronic Conference on Sensors and Applications, 1–15 Nov 2022; Available online: <https://ecsa-9.sciforum.net/>.

Abstract: The natural environment is of the utmost significance not only for a particular location but also for the entire world. This is because the natural environment provides essential environmental services to the human population. However, the environment is being negatively impacted by human activity as well as population growth. The most significant impact was felt in the national capital region. Using the Google Earth Engine (GEE) cloud platform and the QGIS desktop, the purpose of this research is to analyse the changes in land use and land cover (LULC) transformations that have taken place in the Sonipat district of India over the past ten years (2011–2021). Change detection (CD) of a LULC map is a method that examines shifts in LULC throughout time. Landsat 7 and the Sentinel 2 satellite image collections were utilized in this study. The study area was divided into four LULC categories using the most likely classified approach to quantify the changes over the aforementioned period. The results indicated that between 2011 and 2021, cropland in the study area decreased by about 11%. Built-up and the urban area increased by 3%. With the help of this study, decision-makers will be able to make choices that are appropriate in the given situation. The findings emphasize the value of satellite monitoring in reducing the rate of environmental degradation in the Sonipat district.

Keywords: Google Earth Engine; land use land cover; climate change; land degradation; change detection; urbanization

Citation: Rana, D.; Kumari, M.; Kumari, R. Land Use and Land Coverage Analysis with Google Earth Engine and Change Detection in the Sonipat District of the Haryana State in India. *Eng. Proc.* **2022**, *4*, x. <https://doi.org/10.3390/xxxxx>

Academic Editor: Francisco Falcone

Published: 1 November 2022

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1. Introduction

The LULC is undergoing significant development in the vast majority of countries right now [1]. The majority of the blame for these LULC shifts should be placed on humans and the environment in which they live [2]. There will be detrimental effects on human health as well as ecosystems [3]. These changes have had a number of unfavorable effects on a global scale [4], including but not limited to the following: erosion, increased runoff, flooding, the depletion of water resources, and a decline in the quality of the water [5]. The term “land cover” refers to the natural covering that is present on the surface of the land, whereas the term “land use” refers to the activities that are carried out by humans on the land itself [6]. There is a distinction to be made between the two of these terms. The shift in LULC is cause for concern, as it has the potential to have important repercussions for the environment [7] at all different scales (local, regional, national, and global) [8]. The urbanization that is taking place as a direct consequence of the quickening pace of development is having an effect on the changes that are taking place in LULC [9].

In the world we live in today, innovative technologies such as remote sensing (RS) [10] and geographic information systems (GIS) can provide helpful data and tools to assist in the resolution of issues such as these [11,12].

To assess the transformation in LULC between 2011 and 2021, this study examines decade variation in LULC and predicts the shift in LULC that will occur in 2031. This study was conducted on GEE cloud platform and QGIS Desktop 2.18.0 software. With the aid of this study, policymakers will be able to monitor and mitigate the negative effects of LULC change while maintaining the production of essential resources.

2. Study Area

The Haryana district of Sonipat has been chosen as the study area (Figure 1). The latitudes and longitudes of the Sonipat district are 28° 48' 15" and 29° 17' 10" north and 76° 28' 40" and 77° 12' 45" east, respectively. The Survey of India topo sheets 53C, 53D, 53G, and 53H cover an area of 2213.37 km². Panipat is to the north of the district, Jind is to the west, Rohtak is to the south-southwest, and Delhi is to the south. In addition, the city is connected to Delhi and Chandigarh by broad-gauge railway. The district experiences intense heat in the summer and extreme cold in the winter. From late November to March, we experience the cold season. The southwest monsoon season occurs between July and September.

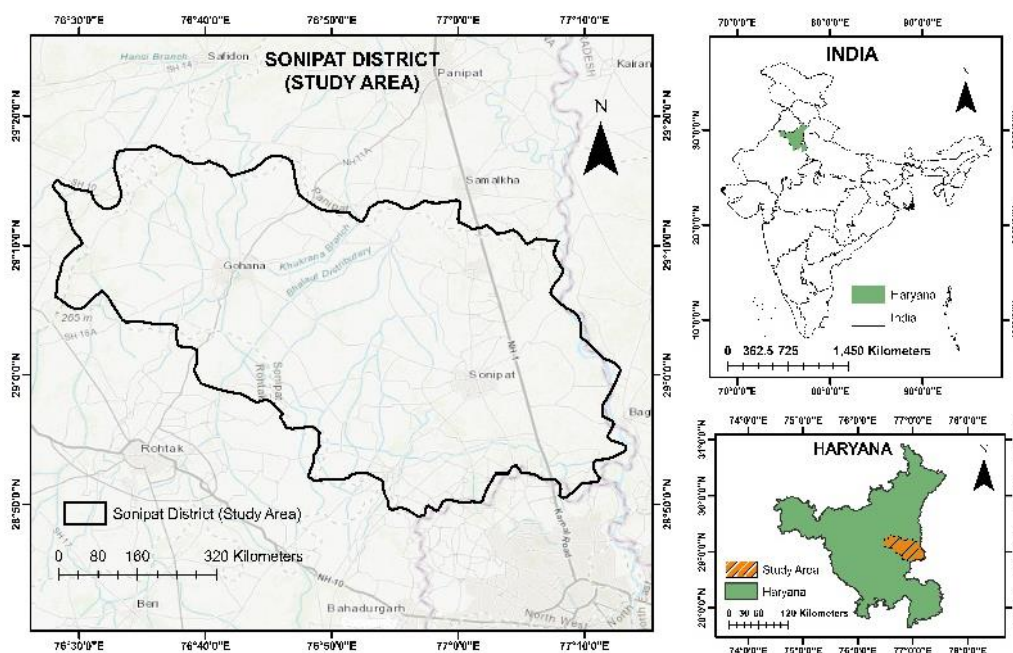


Figure 1. The geographic location of the area of study.

3. Methodology and Data Used

3.1. Land Use Land Cover Classification

The LULC studies made use of data from both Landsat 7 (2011) and Sentinel 2 (2021). The following major LULC classes were chosen for this research: urban and built-up areas, bodies of water, cropland, and barren/fallow areas (Other). Using GEE script, LULC was categorised. This cloud-based platform implements the three major steps outlined below:

- Image Selection.
- Collection of training samples.
- Run the classifier.

3.2. Accuracy Assessment

The accuracy of both LULC classification maps is evaluated with the help of a confusion matrix that has been built into GEE. This matrix does this by contrasting the LULC that is linked to the validation points with the classifications that are produced (2011, 2021). The overall accuracy can be calculated with the help of a confusion matrix.

3.3. Change Detection and Prediction

Using the cellular (MOLUSCE) plugin of Quantum 283 GIS 2.18.0, both the detection and prediction of LULC changes were identified. A change in area is computed between the initial year (2011) of the LULC and the final year (2021) of the LULC. Using the transition potential model, 2031 predictions were made.

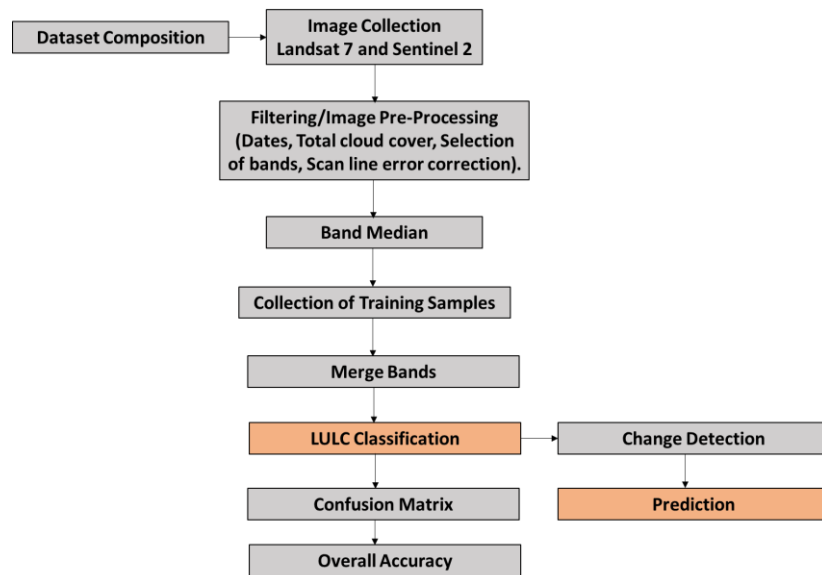


Figure 2. Methodology flow chart.

4. Results and Discussions

4.1. LULC Classification

As was just stated, estimates for the LULC classification were done for the years 2011 and 2021. In 2011, the majority of the land was used for agricultural purposes; however, by 2021, large swaths of land had been developed into urban and built-up areas, particularly in the south and south east regions of the district (Figure 3).

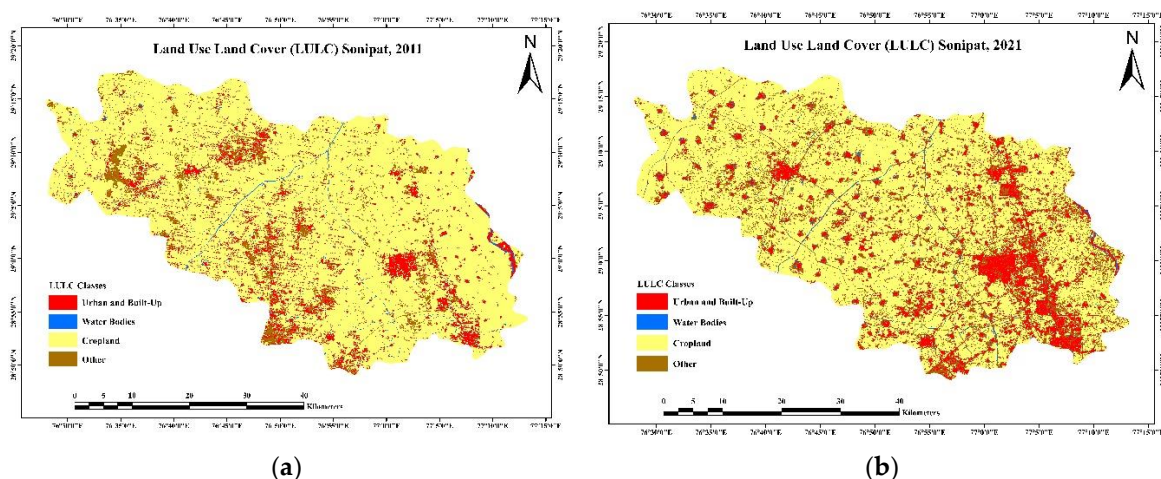


Figure 3. Sonipat district LULC classification: (a) 2011 classification; (b) 2021 classification.

4.2. Accuracy Assessment

The accuracy of the classification was analyzed with the help of a confusion matrix. The accuracy of the classification was 0.98 over the course of both years. Same accuracy techniques are used by other researchers in their studies [13].

Table 1. Evaluation of the accuracy assessment of the LULC classification for the year 2011.

Validation Error Matrix 2011				
LULC Classes	4 Elements			
Urban and Built-Up	16	0	0	0
Water Bodies	0	15	0	0
Cropland	0	0	14	0
Other (Fallow land and Barren land)	1	0	0	14
Validation overall accuracy	0.98			

Table 2. Evaluation of the accuracy assessment of the LULC classification for the year 2021.

Validation Error Matrix 2021				
LULC Classes	4 Elements			
Urban and Built-Up	30	0	0	1
Water Bodies	0	14	0	0
Cropland	0	0	18	0
Other (Fallow land and Barren land)	0	0	0	22
Validation overall accuracy	0.98			

4.3. Change Detection

Changes in LULC are analyzed as a proportion of total land area. Positive values indicate improved categorizations, whereas negative values indicate categorizations that have deteriorated. Urban and built-up LULC class area is observed to increase by 3 percent in 2021, while other LULC classes, includes fallow and barren land, are also observed to increase by 7 percent in 2021 (Table 3). These LULC changes are observed in the study area as a result of its incorporation into the national capital region, so these changes are the result of human activity and migration.

Table 3. LULC change analysis for the years 2011 and 2021.

LULC Classes	2011 Area km ²	2021 Area km ²	Δ	2011%	2021%	Δ %
Urban and Built-Up	234.81	319.87	85.06	10.60	14.45	3
Water Bodies	30.55	22.81	-7.74	1.38	1.35	-0.34
Cropland	1843.61	1595.11	-248.49	83.29	71.41	-11
Other (Fallow land and Barren land)	112.14	267.84	155.69	5.06	12.79	7

Figure 4 depicts the trends of the change in LULC from 2011 to 2021. These trends indicate that the cropland land use class covered a greater proportion of the total land area in the study area, whereas the waterbodies LULC class covered a smaller area. Agriculture is the primary source of income for the people who live in this district of Haryana, and the Khadar region and the upland plains of this district are ideal for the cultivation of rice and sugarcane, respectively.

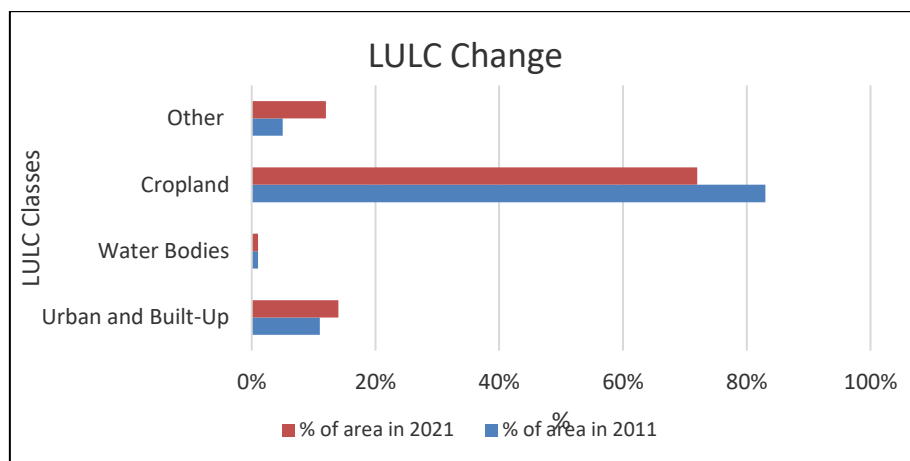


Figure 4. LULC change trends for 2011 and 2021.

4.4. Prediction

As mentioned earlier, a prediction was made for the year 2031 by using the QGIS MOLUSCE plugin. The results of this prediction are presented in Table 4 and Figure 5. Table 4 demonstrates how the LULC classes have evolved from 2021 to 2031. An increase in area can be seen in urban, built-up and other LULC class (5.04%, 3.87). Both the area of cropland and water bodies is expected to decline in the coming years. The extended LULC classes are represented by a variety of colors in Figure 5.

Table 4. Analysis of the LULC forecast for 2031 based on the LULC maps of 2011 and 2021.

LULC Classes	2021 Area km ²	2031 Area km ²	Change km ²	2021%	2031%	Δ %
Urban and Built-Up	319.87	431.54	111.67	14.45	19.49	5.04
Water Bodies	22.81	14.95	-7.86	1.35	0.69	-0.66
Cropland	1595.11	1398.04	-328.52	71.41	63.16	-8.25
Other (Fallow land and Barren land)	267.84	368.84	101	12.79	16.66	3.87

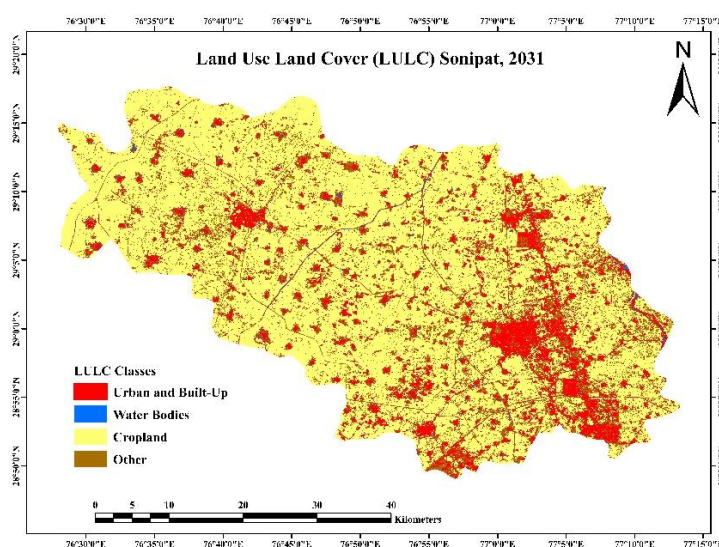


Figure 5. Forecasted LULC map for the year 2031.

5. Conclusions

Predictions of LULC are absolutely necessary for the creation of plans that are capable of striking a balance between efforts to conserve resources, competing user needs, and

the pressures of development. To simulate and make predictions regarding the future LULC maps for the Sonipat district, the MOLUSCE plugin is utilized. Because of urbanization, the proportion of land that is classified as urban and built up within the LULC is increased over the three decades (2011, 2021, and 2031) while the proportion of cropland is decreased as a result of the influence of the capital region within the study area.

Author Contributions:

Funding:

Institutional Review Board Statement:

Informed Consent Statement:

Data Availability Statement:

Conflicts of Interest:

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