





Land Degradation (UN SDG 15.3.1) and Urban Expansion in Lucknow Using MODIS and Landsat (2001-2020)

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Problem & Motivation

- ♦ Land degradation + rapid urban expansion = critical challenges for SDG 15.3.1.
- ♦ Lucknow, India: population ↑ from $2.3M (2001) \rightarrow 3.7M (2020)$.
- ♦ Need reproducible, open workflow to monitor land change.



Fig. 1: Location of Lucknow within Uttar Pradesh, India.

Data Sources

- ♦ MODIS MCD12Q1 (500m, annual, IGBP classes).
- ♦ Landsat 7/8 (30m, NDVI composites).
- Study Area: Lucknow, Uttar Pradesh.
- ♦ Time frame: 2001 vs. 2020.

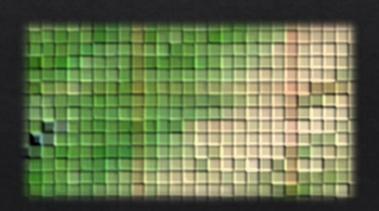




Fig. 2: Area of interest (From GEE).

Workflow / Methodology

Built entirely on Google Earth Engine. Steps involved:

- ♦ MODIS land-cover pixel counts (2001, 2020).
- ♦ Landsat NDVI composites & statistics.
- ♦ Merge + sample 1,000 points for NDVI—land-cover analysis.
- ♦ Validation with zoning maps & high-res imagery.
- ♦ Open-Source Code: https://github.com/rizvizahra/lucknow_sdg_15_3_1/

Algorithm 1 Land Cover and NDVI Analysis in Lucknow

- 1: Define Study Area: Set geographic bounding box for Lucknow (Lat: 26.72–26.96, Lon: 80.80–81.10)
- 2: Select Years of Interest: Define years to compare
- 3: Load MODIS Land Cover Data: Import MODIS MCD12Q1 product and extract IGBP classes
- 4: Load and Preprocess Landsat Data:
 - Import Landsat 7 and 8 collections
 - · Apply cloud and shadow masking
 - Normalize reflectance and compute NDVI
 - Generate annual median composites
- 5: Reproject and Merge Datasets: Align Landsat and MODIS resolutions (500 m) and combine NDVI with
- 6: Generate Sample Data: Randomly sample 1000 pixels 7: Visualize Results:

 - Plot histograms of land cover class distributions
 - Create bar charts comparing pixel counts per class • Display NDVI histogram and scatter plot of NDVI vs.
- 8: Compute Summary Tables:

 - Count pixels per IGBP class for each year
- Compute NDVI statistics (mean, min, max) per year 9: Export Results: Print land cover and NDVI statistics

Results: Land Cover Change

- ♦ Cropland ↓ 13.9% (Class 12).
- ♦ Urban ↑ 17% (Class 13).
- ♦ Grassland + Savanna ↑ in peri-urban zones.

Table I: Land–cover class counts for selected classes.

Class	Description	Count 2001	Count 2020	Δ
9	Savanna	4	52	+48
10	Grassland	24	169	+145
12	Croplands	2,357.77	2,029.85	-327.92
13	Urban	791.31	925.31	+134.00
14	Cropland	0	0.92	+0.92

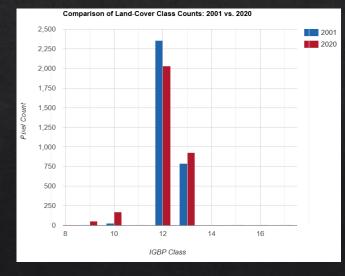


Fig. 3: Comparison of IGBP land—cover class pixel counts between 2001 and 2020, based on MODIS MCD12Q1 data.

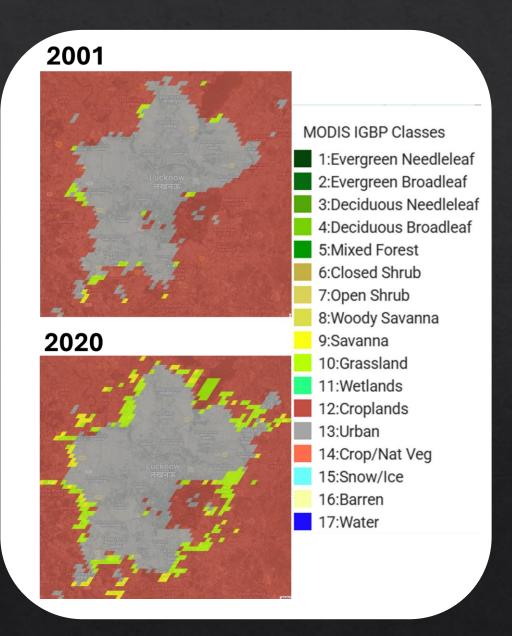


Fig. 4: Expansion of the urban area over time.

Results: Vegetation & ML Pilot

- \diamond NDVI mean declined by -0.0315 (weaker vegetation health).
- ♦ High NDVI patches = irrigated fields, parks.
- ♦ Logistic regression (proof of concept): 82% accuracy (urban vs cropland vs grassland).

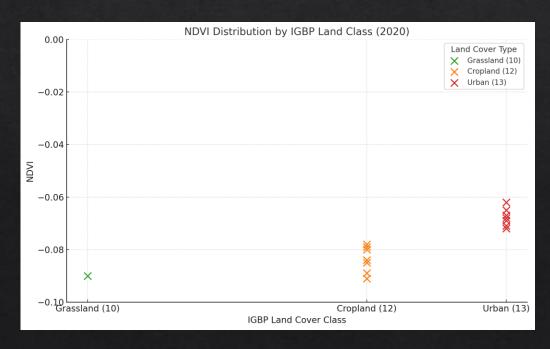


Fig. 5: Scatter of NDVI vs. IGBP Class (2020).



Fig. 6: Confusion matrix for MODIS land-cover classification (2020), derived from stratified validation using 300 visually verified reference points.

Impact & Future Directions

- ♦ Supports urban planners & humanitarian practice (heat stress, food security).
- * Reproducible, open workflow for SDG monitoring.
- ♦ Future: multi-seasonal composites + socioeconomic integration + ML classifiers.
- ♦ Public repo link https://github.com/rizvizahra/lucknow-sdg-15-3-1/



References:

- 1. NASA LP DAAC. MODIS Land Cover Type (MCD12Q1) Version 6 [Data set]. NASA EOSDIS Land Processes DAAC. https://doi.org/10.5067/MODIS/MCD12Q1.006
- 2. U.S. Geological Survey (USGS). Landsat Collection 2 Level-2 Surface Reflectance Products [Data set]. https://doi.org/10.5066/P9C7I13B
- 3. Google Earth Engine. A planetary-scale platform for Earth science data & analysis. https://earthengine.google.com/