

**Geospatial Data Analysis (CS321)**

# **Forest Cut Temporal Detection**

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# Deforestation

Deforestation is The large-scale clearing or thinning of forests, mainly due to human activities like logging, agriculture, and urbanization.

It leads to loss of biodiversity, disrupts ecosystems, contributes to climate change, and increases risks of natural disasters like floods and soil erosion.

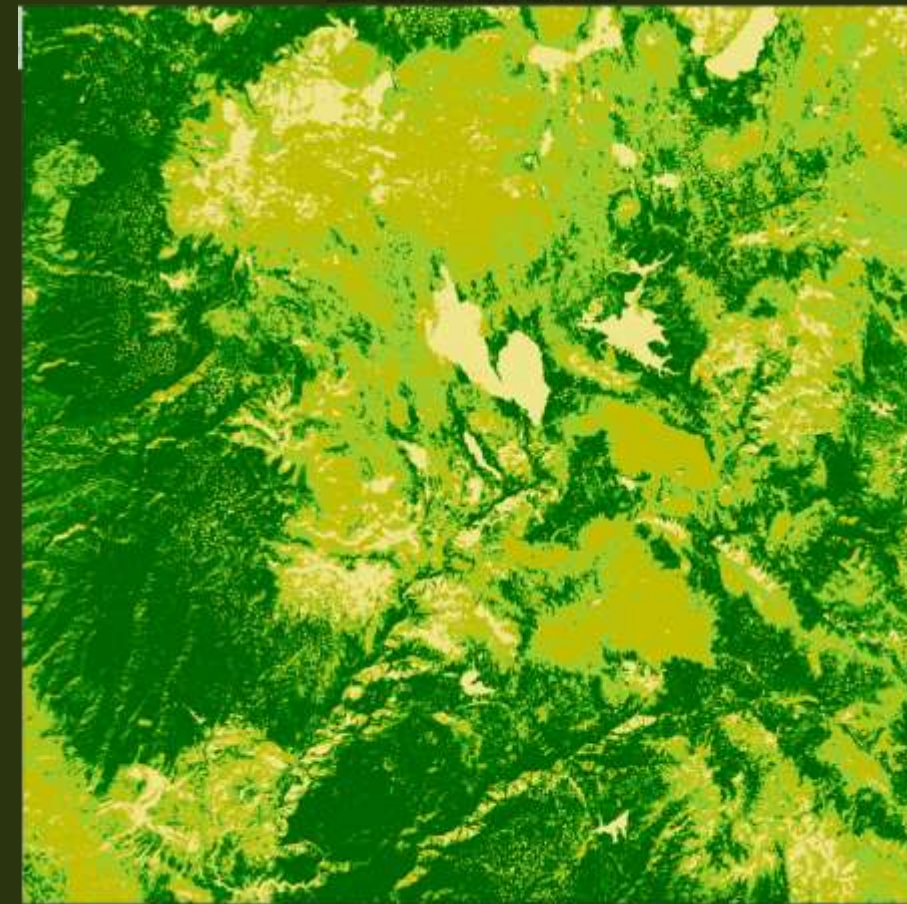
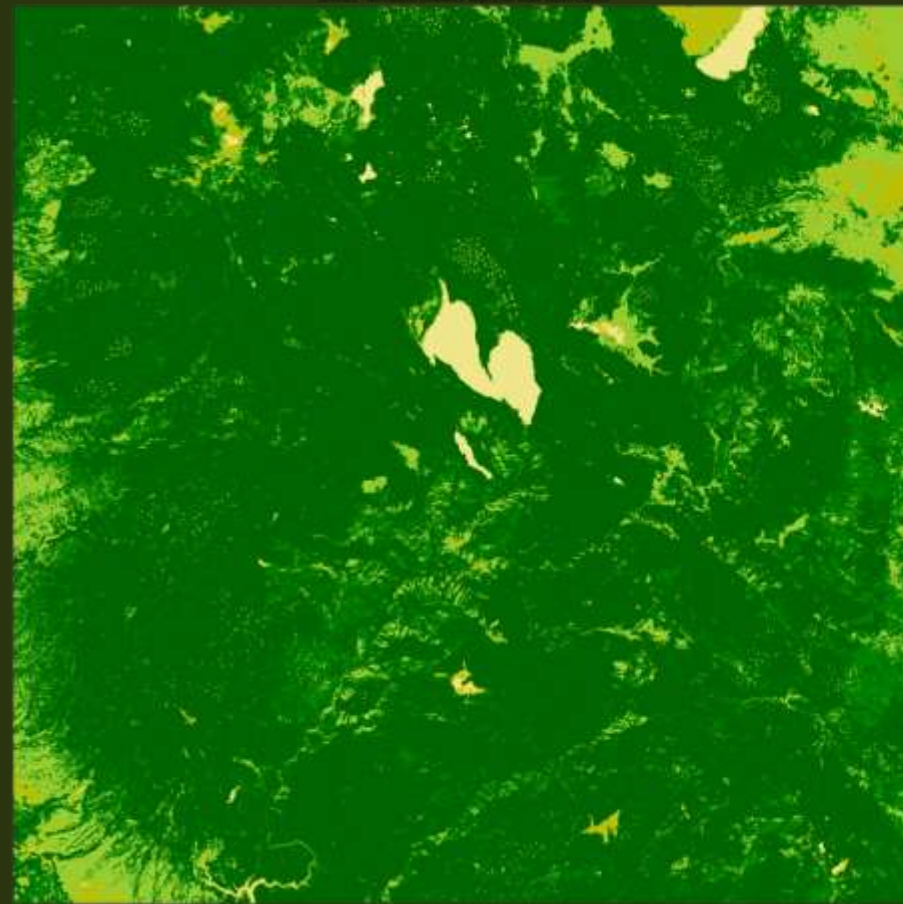
# Introduction



- Deforestation, driven by human activities and natural events, devastates ecosystems, impacts climate patterns, and accelerates global warming and biodiversity loss.
  - Rising deforestation rates threaten environmental sustainability, with forest fires and logging disrupting landscapes.
  - Detecting and monitoring forest cover changes over time can aid conservation efforts, helping identify affected regions and implement timely interventions.
  - Using satellite imagery, this project analyzes temporal forest cover changes, revealing deforestation patterns across landscapes.
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# Objective



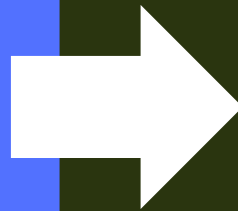
Develop a method to accurately identify and visualize deforested areas over time using satellite imagery to assess forest loss. —→

Calculate and analyze the rate of forest cover change across time intervals to understand deforestation trends and patterns.

# Appraoch

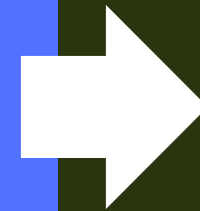
## Setting Up the Environment

- Setting up the environment in **SageMaker** Studio Lab, selecting or uploading a **YAML** configuration file.
- Once configured, either install packages manually or import them to manage and analyze geospatial data effectively.



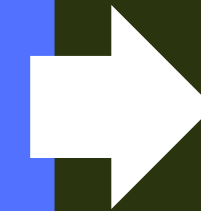
## Data Retrieval and Preparation

- **Sentinel-2** geospatial data is accessed from the **AWS** registry to analyze deforestation using temporal detection.
- Define search coordinates, choose cloud-free tiles, and set a time window, optimizing for minimal cloud cover to ensure accurate data analysis.



## Spectral Analysis with Sentinel-2

- Download specific spectral bands, like **visible**, **NIR**, and **SWIR**, to calculate indices such as **NDVI**.
- Use these indices to distinguish between healthy and damaged vegetation, providing insights into changes across the observed area.



## Visualizing Vegetation Indices

- Utilize **earthpy** to plot NDVI indices, classifying the spectral data into bins.
- Through visualization, identify vegetation conditions over time, distinguishing areas with significant deforestation.



# References we are going through..

How people are approaching in literature on all kinds of dataset

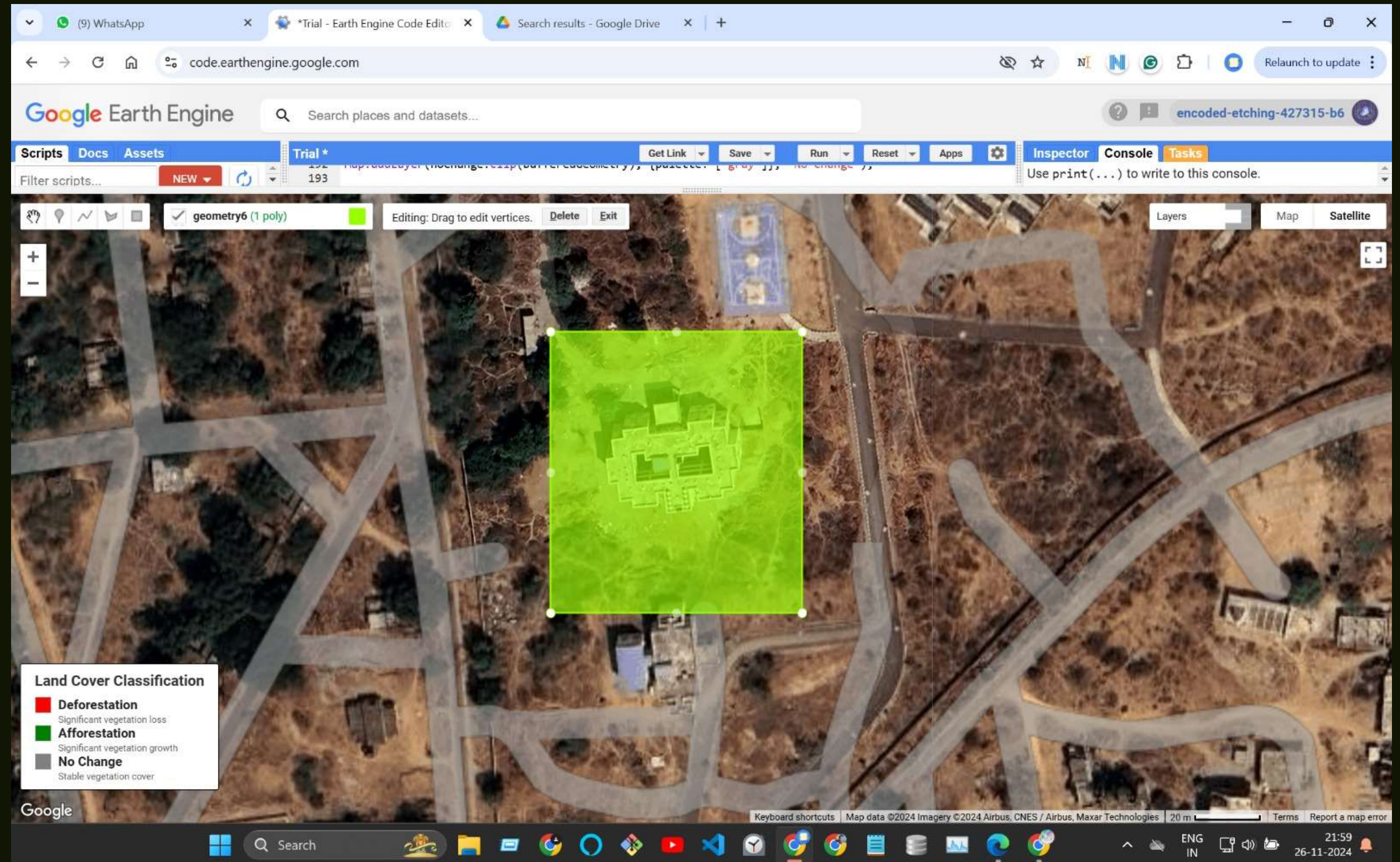
How to use AWS SageMaker to retrieve data from Sentinel Hub and build proof of concepts for detailed analysis of any location.



# Working and Demonstration

## Step-1

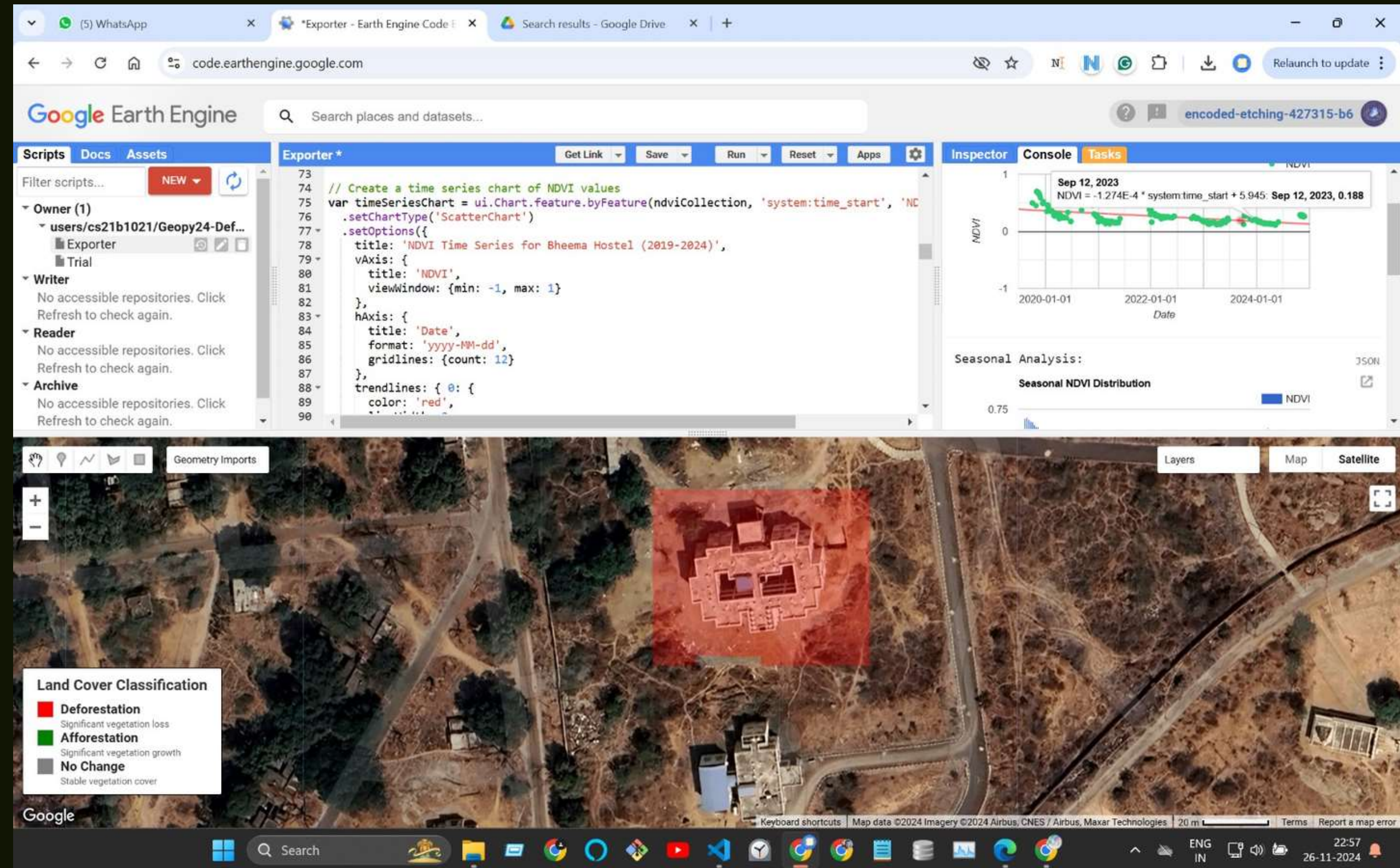
Defining the study area





# Step-2

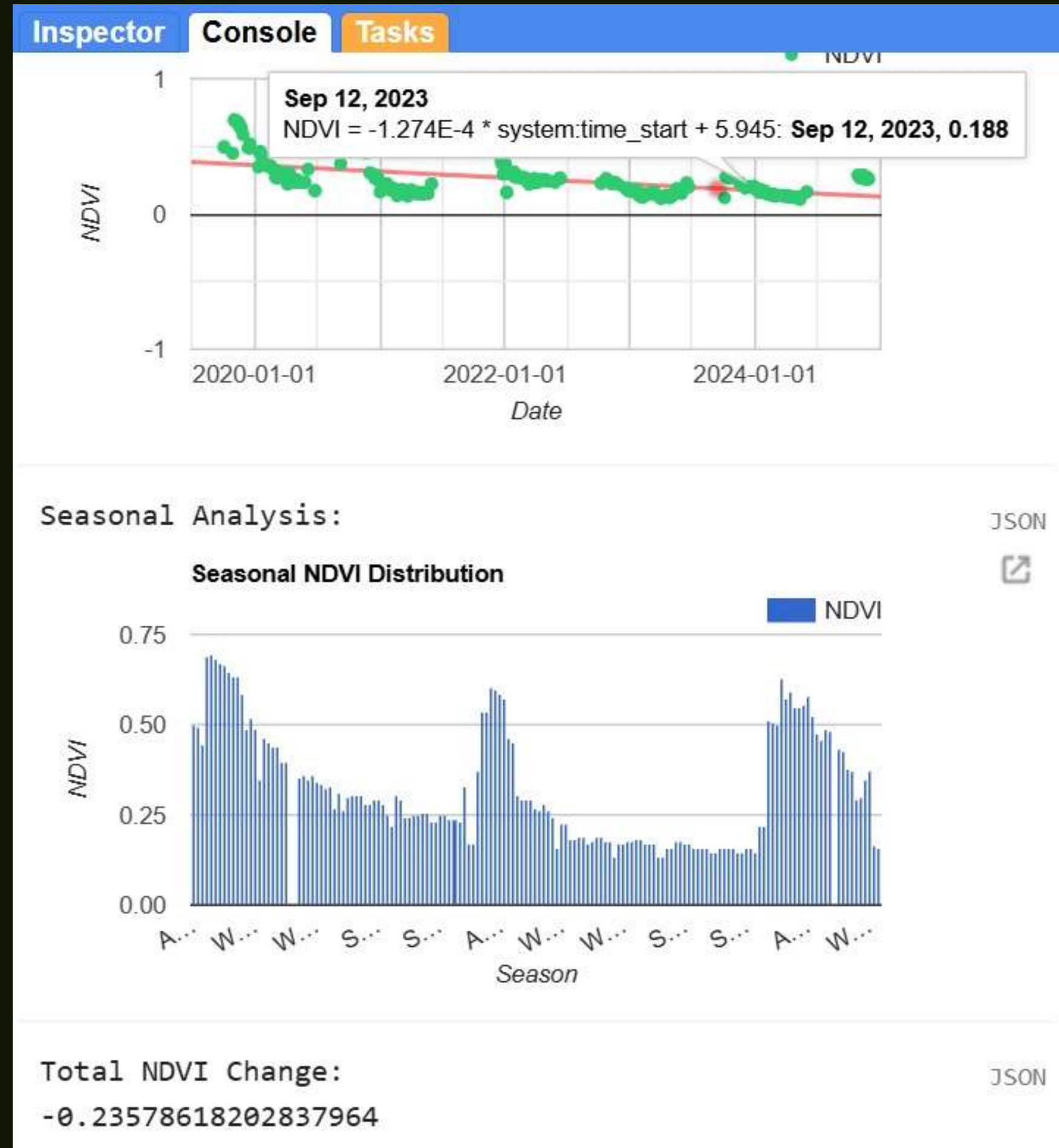
A visualization showing areas categorized into deforestation, afforestation, and stable zones based on NDVI change analysis.



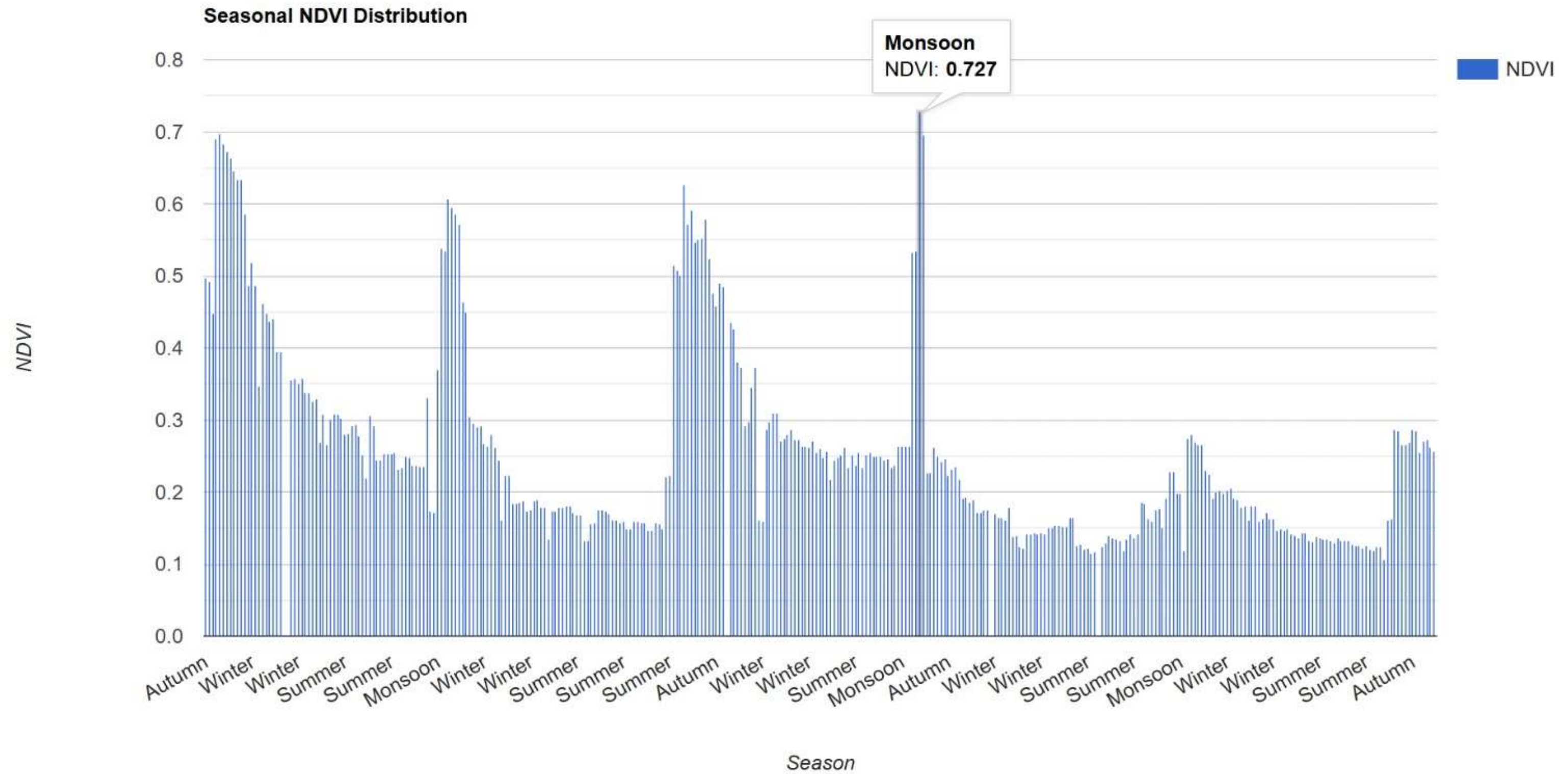


## Step-3

These visualizations provide insights into NDVI distribution and trends, enhancing the understanding of vegetation health changes.





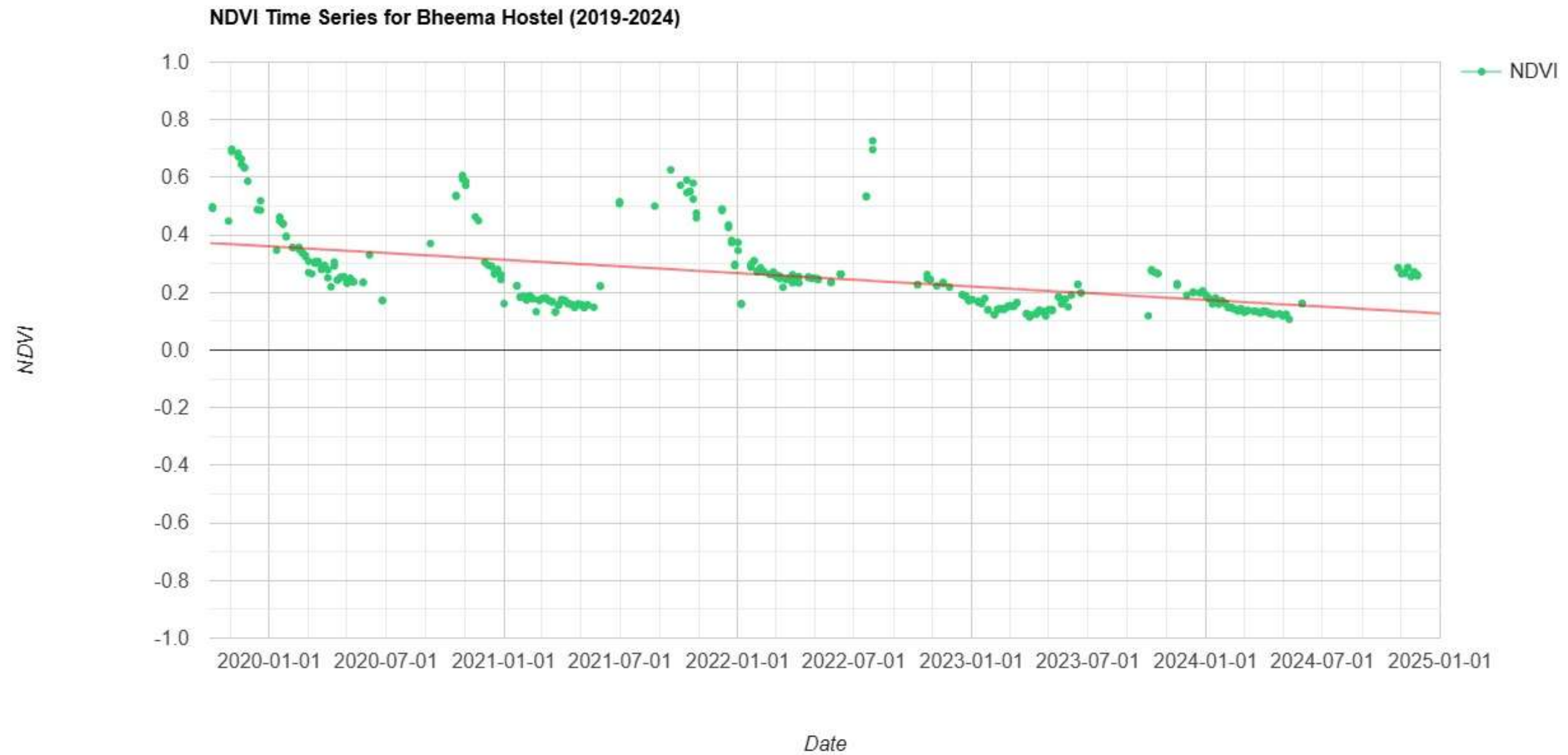


The peak NDVI before construction (July 30, 2022) was 0.726, indicating dense vegetation



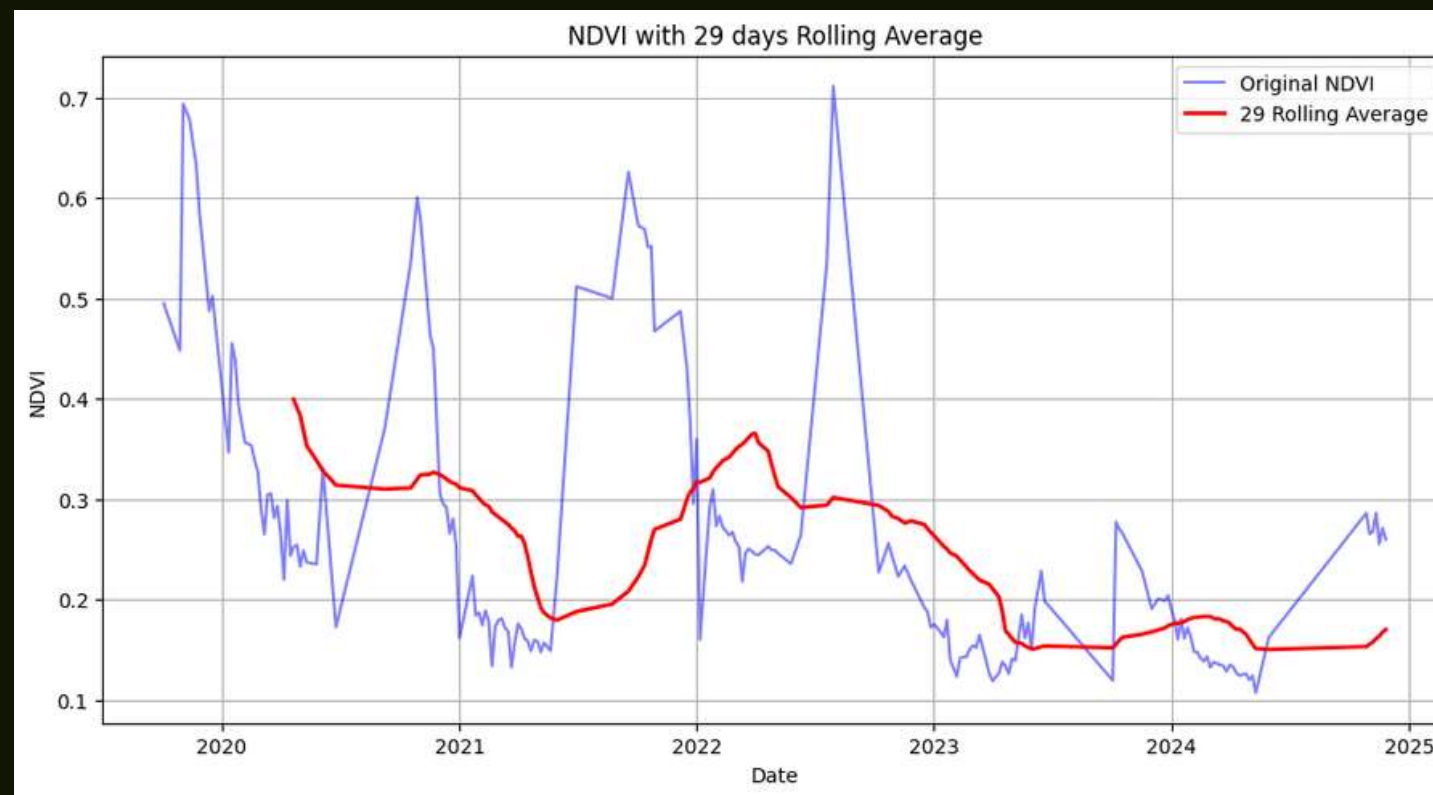
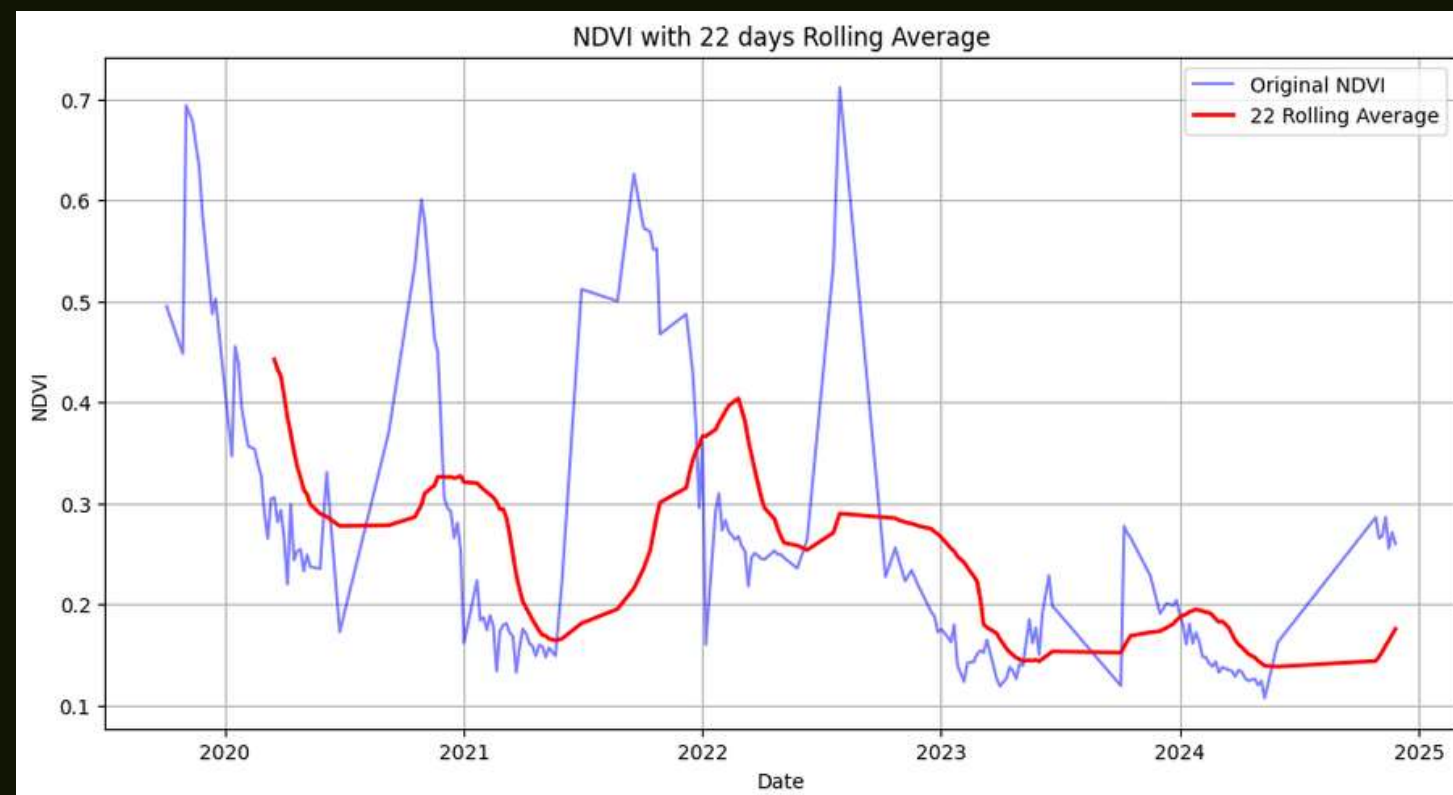
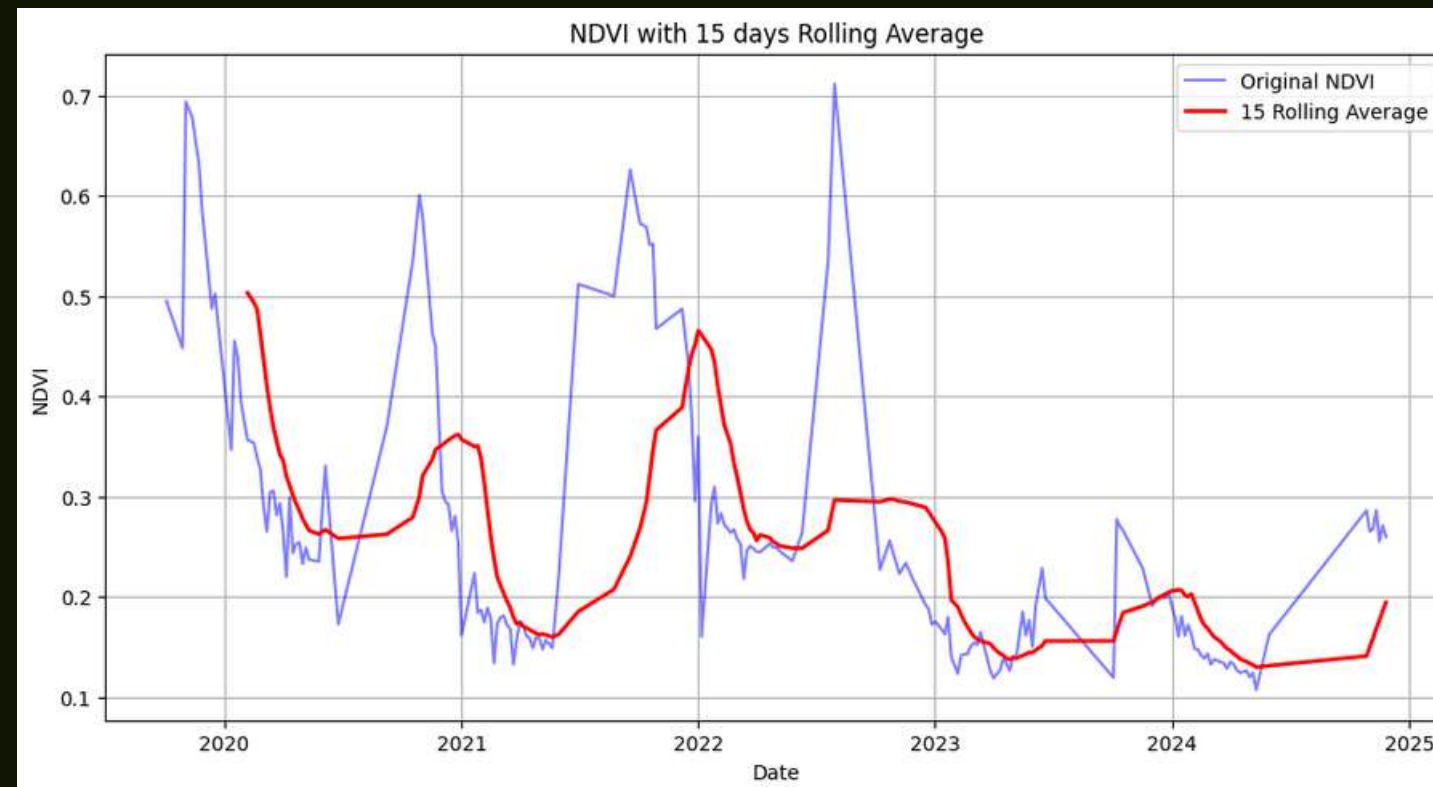
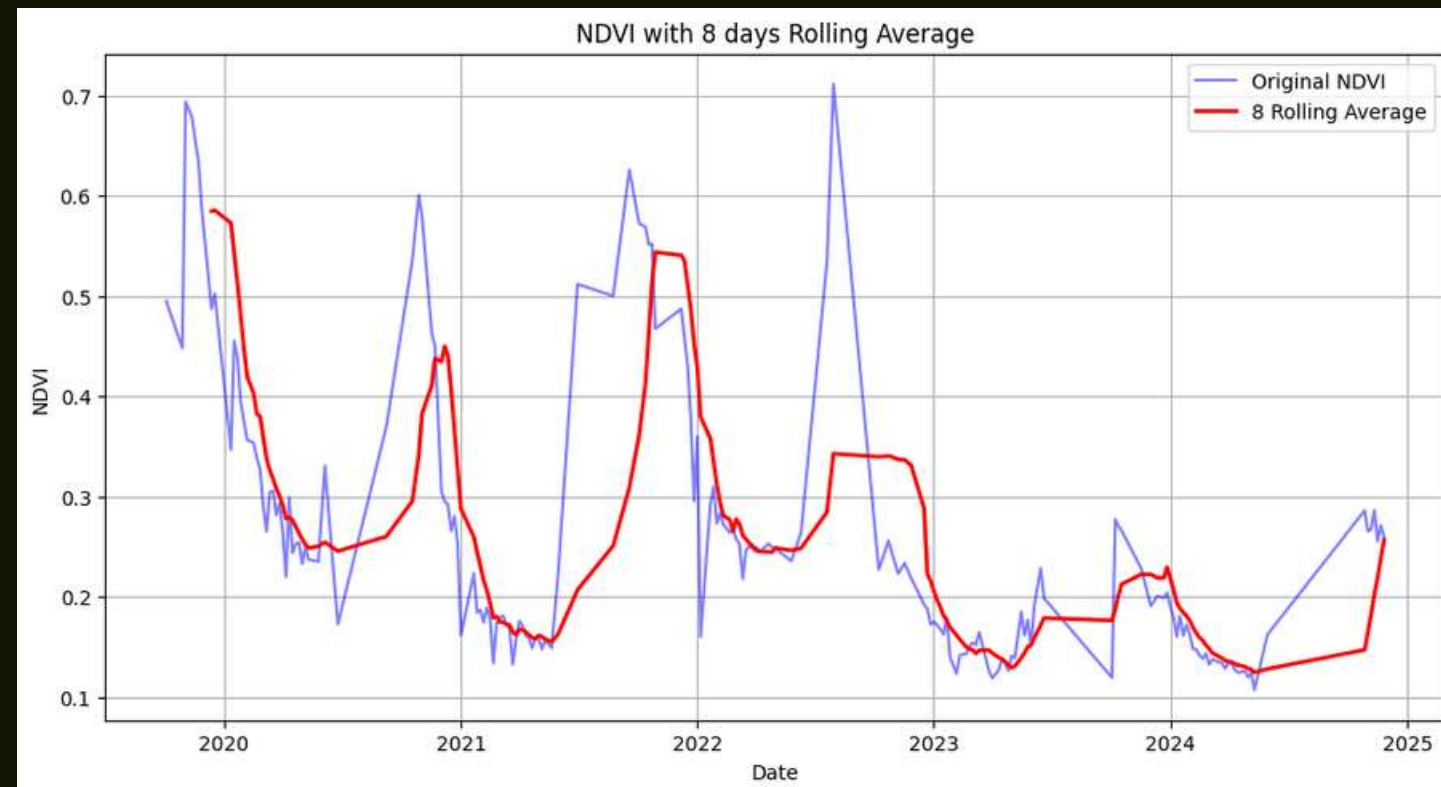






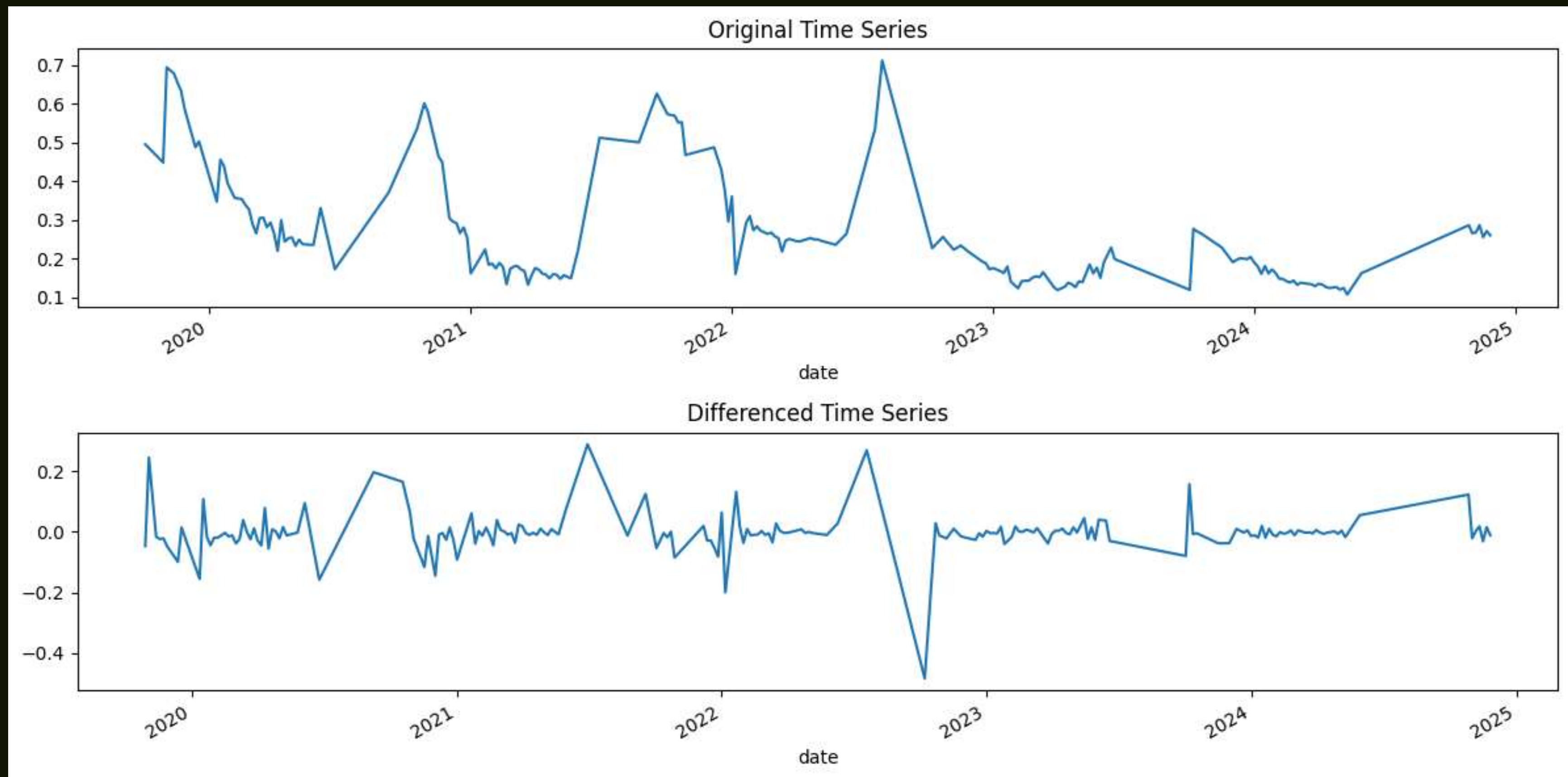
This chart depicts vegetation health trends over the study period, highlighting seasonal variations and long-term changes.





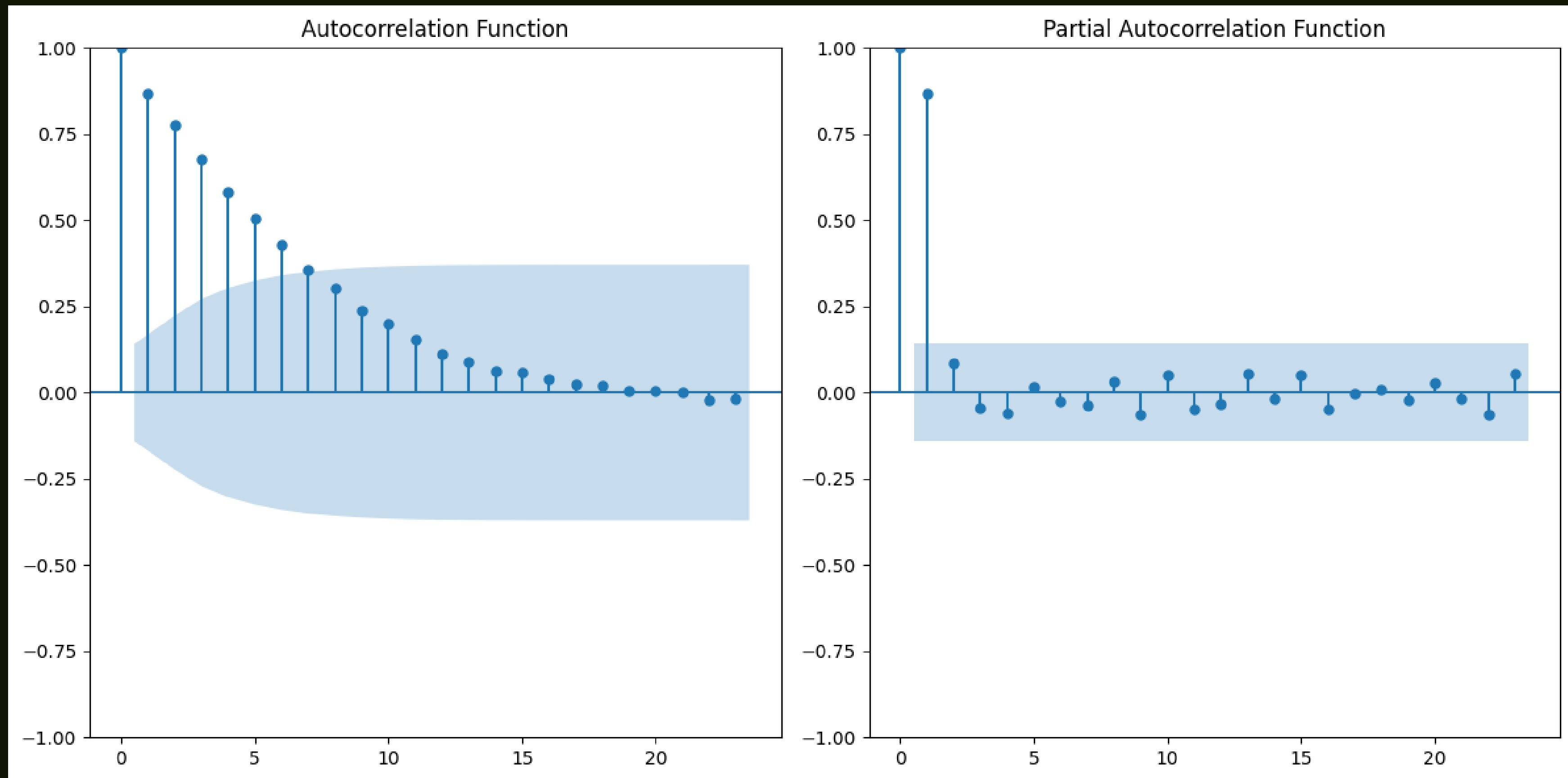
Comparison of rolling averages with window sizes of 8, 16, 22, and 29 days. Rolling averages help smooth noisy data, revealing clearer trends. Among these, the 8-day and 14-day rolling averages perform best, striking a balance between smoothness and fidelity to the data. Larger window sizes tend to over-smooth and obscure finer details in the dataset





Transformation of the dataset into a stationary time series through differencing: Stationarity ensures that statistical properties like mean, variance, and autocorrelation remain consistent over time, which is crucial for reliable modeling and forecasting in time series analysis

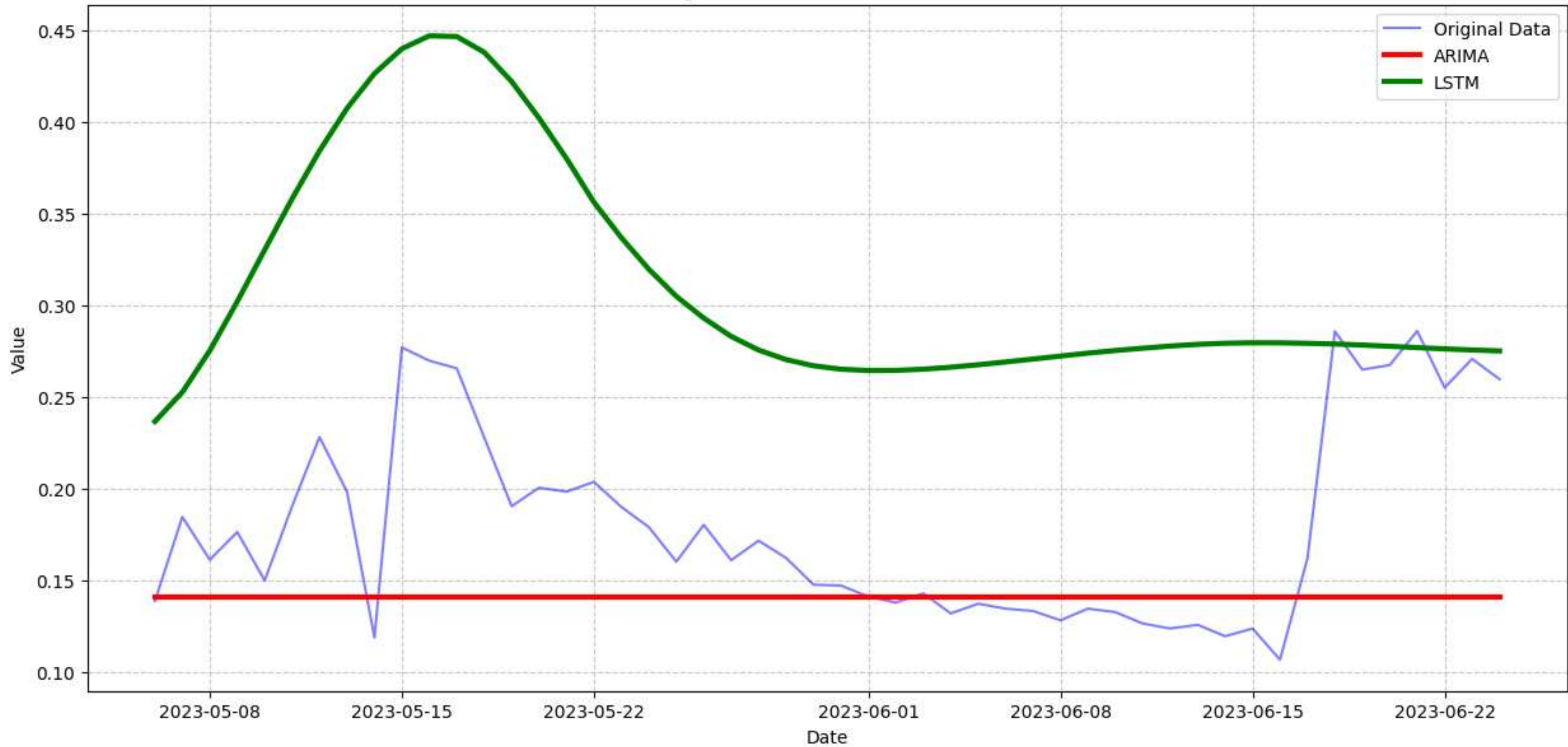




Autocorrelation helps identify the seasonality hyperparameter  $m$  for optimal results. A greater distance from the zero line indicates stronger correlation. Partial autocorrelation typically exhibits an exponential decay, with the highest correlations observed at lags 2 and 3 (it is always 1 at lag 0 by default )



Original Data vs Predicted Values



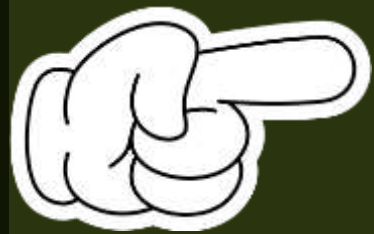
LSTM fits the data and its shape better than ARIMA across some stretch of testing data



# Future Scope



**Real-time Deforestation Alerts:** Advanced technology will enable real-time alerts for immediate response to forest loss.



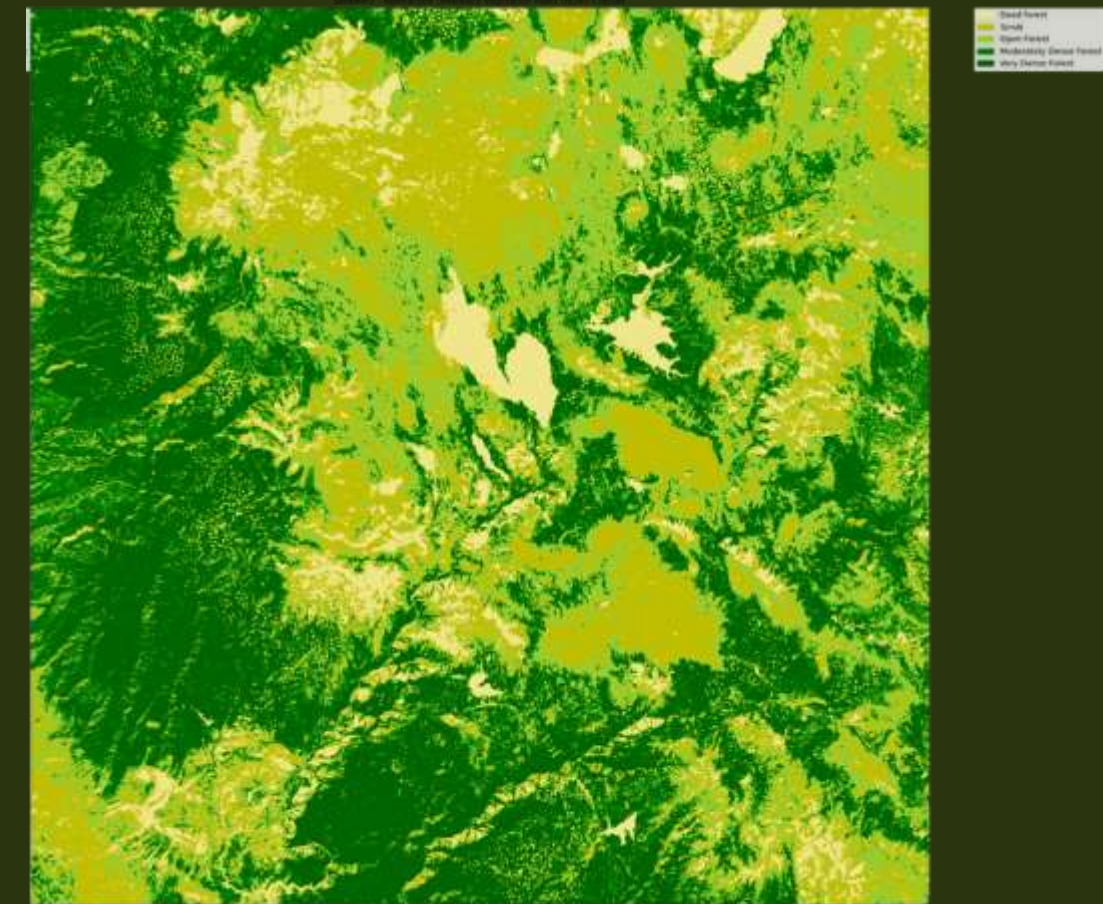
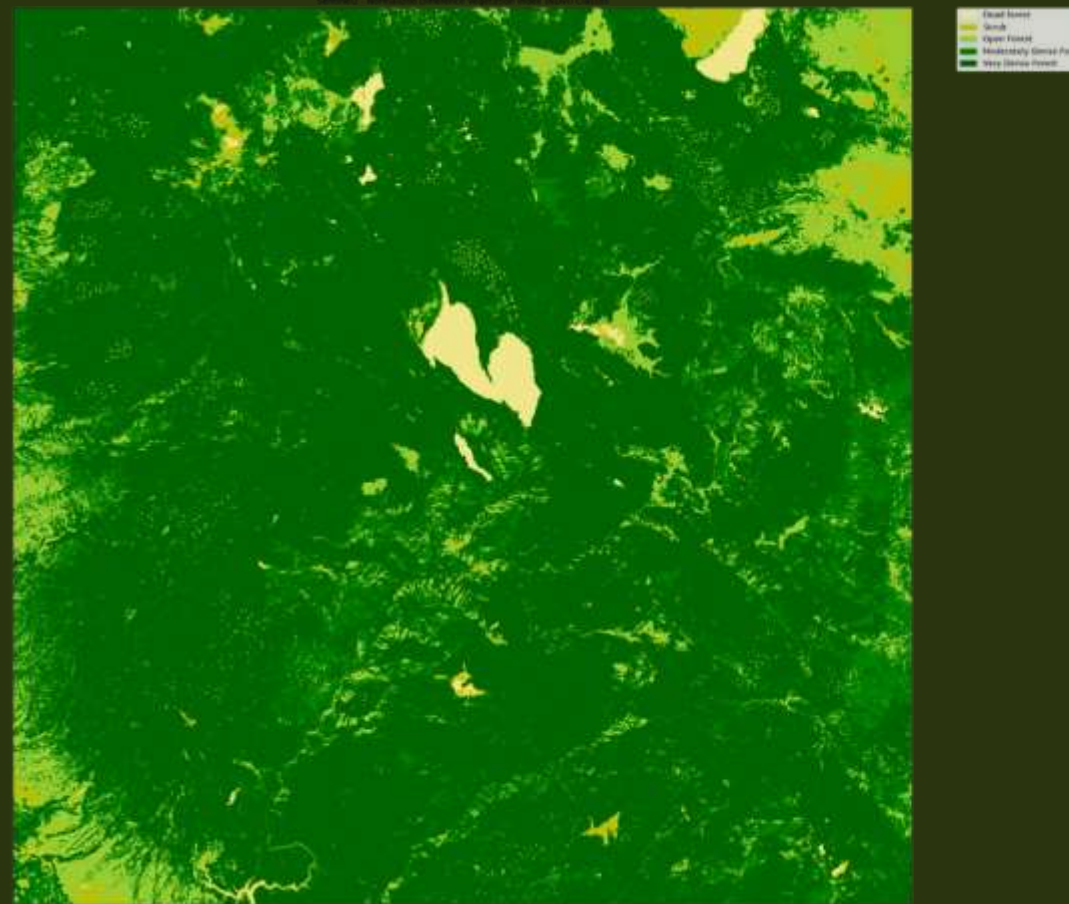
**Support for Conservation Policies:** Accurate data can inform policies, helping governments protect forests and reduce carbon emissions.

**Thank You**



# Forest Cut Temporal Detection

- Objective



Develop a method to accurately identify and visualize deforested areas over time using satellite imagery to assess forest loss accounting for existing vegetation spread.



Calculate and analyze the rate of forest cover change across time intervals to understand deforestation trends and patterns.