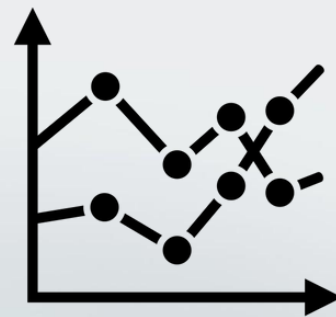




Shrujal Ambati
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Aditya



Data Science

Project objective



Working with Axians' Location Intelligence Team, we developed algorithms using **satellite images** to determine **whether vegetation growth will interfere with power line construction**. This will involve using machine learning to identify and monitor power lines, geographical features, obstructions and other changes in the landscape. This project will involve a lot of **Random Forest** based ensemble methods and segmentation to do **pixel-based classification**.

Geographical Location

Ioana

- The Portuguese Archipelago of Azores
- São Miguel Island
- The surface of the island is 760 km²
- Area of study - denser vegetation that can cause power outages



Steps



Farzad

1

Dataset - Sentinel-2 SR- 2A

2

Cloud Masking

3

Time Series Analysis

4

Random Forest Image
Classification

Cloud Cover

Aditya

- Clouds covered a significant portion of our area of interest
- Need to remove before land analysis
- Many problems with existing cloud masking algorithms
- Solution: sentinel-2 cloudless algorithm



Vegetation Management

Aditya

- The detection of vegetation cover changes through the creation of vegetation indices has become an important application of remotely sensed imagery.
- Remotely sensed information from Landsat, SPOT, MODIS and NOAA satellites are a valuable tool for graphically understanding changes in vegetation and land use at a range of temporal and spatial scales.

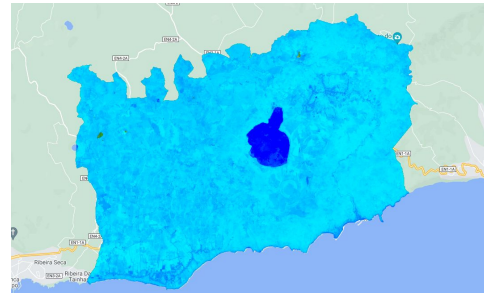
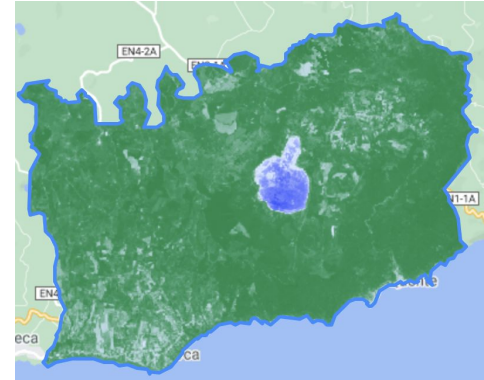


NDVI and other indices

Ioana

- NDVI is the most commonly applied index
- NDVI is a proxy of primary productivity and is directly related to plant growth and density
- Vegetation indices produce a single image indicating the amount of green vegetation production and biomass
- Low index values usually indicate less vegetation while high values indicate higher photosynthetic activity.
- Other spectral indices: NDWI, EVI, ARVI, SAVI

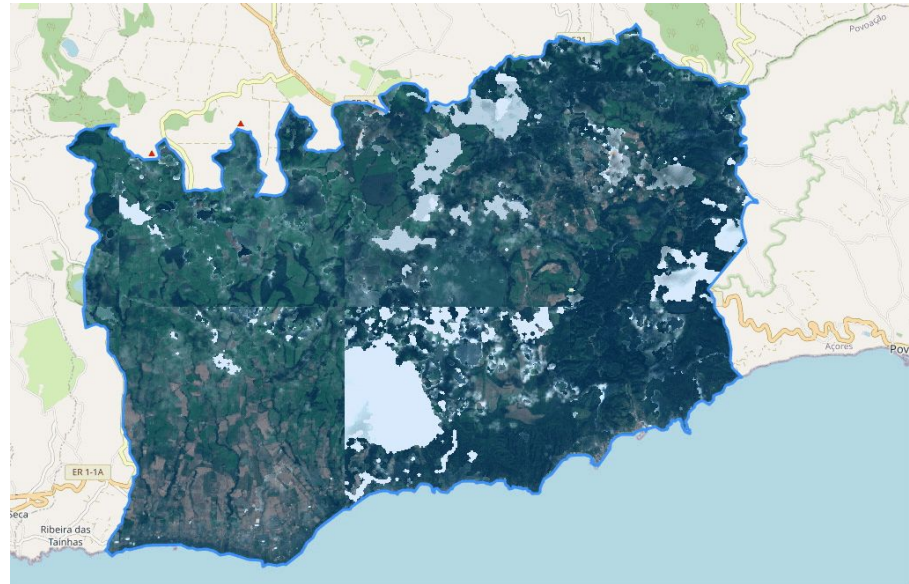
$$NDVI = \frac{(\text{Near Infrared} - \text{Red})}{(\text{Near Infrared} + \text{Red})}$$



Monthly composites

Farzad

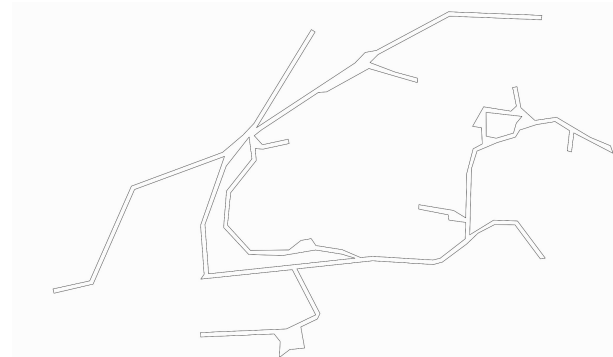
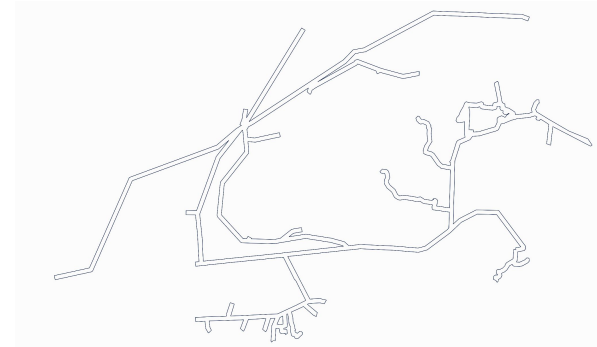
- Sentinel-2 satellite has a 10 days temporal resolution
- Calculated median images
- This resulted in diminished cloud cover and better quality images
- Compared the results with daily images, as well as biweekly and quarterly composites



Working with Geospatial data

Shrujal

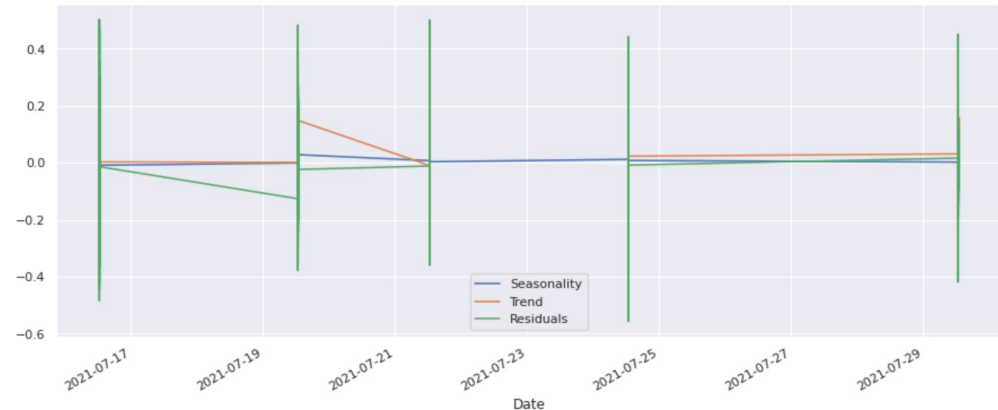
- The vector format indicates the location and shape of features and boundaries accurately.
- Point, line, shape, polygon or area
- E.g. shapefile (.cpg, .dbf, .prj, .shp, .shx), kml, or geoJson format
- Using the Earth Engine API we loaded our powerline corridor in all these different formats, however the error was with collabs capacity
 - Possible solution using image reduction (reduce complexity of the multipolygon)



Time Series visualizations

Shrujal

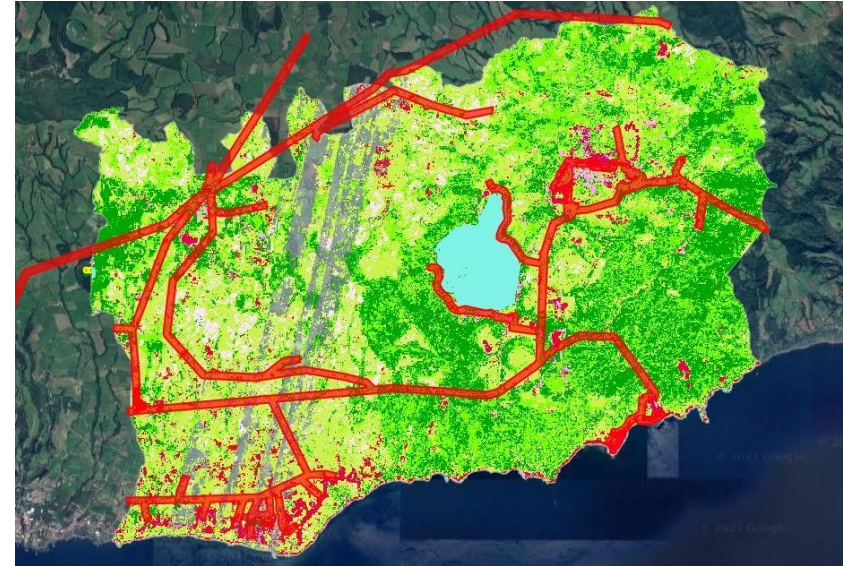
- Started by using the sentinel-2 cloudless collection using a function in python applied to all the images
- Tried using geemap, geopandas, and functions from the earth engine api to convert a shapefile of the power line corridors to a geojson format or just a plain polygon
 - Issue was size and complexity of the corridors
- To see if it would work tried using a point in the study area with a large buffer and converted that region to information I could use to
- calculate NDVI from the bands and then put the NDVI information into a pandas dataframe which was successful, and we saw a plot of NDVI over time which also looked reasonable



Random Forest Classification

Shrujal

1. Importing and Visualizing Data in GEE
 - a. Import Study Area
 - b. Import training points and validation points
 - c. Cloud masking with sentinel 2
2. Setting up and implementing classification
 - a. select the spectral bands that are relevant for the classification
 - b. The training points will be used to extract the reflectance values of the pixels from all spectral bands and this will be passed to the classifier algorithms.
 - c. Used Copernicus CORINE Land Cover catalog to catalog pixels classified by color
 - d. CART and RF



overall accuracy: CART **1.000**, RF **0.968** for Date Range: 2021-01-01 → 2021-07-01



Future Steps

- Use our different composites as testing data to test accuracy of random forest model
- Apply prediction algorithm to predict vegetation encroachment on power lines, in order to advise current and future construction
- Organize all of our past work in a way that fits industrial standards

What we learned

Farzad

- Geospatial Data Files and Formats
- Machine Learning Algorithms
- Google Earth Engine
- Image Classification
- Cloud Masking
- Teamwork, Organization and working across timezone



Aditya



Meet The Team



Shrujal



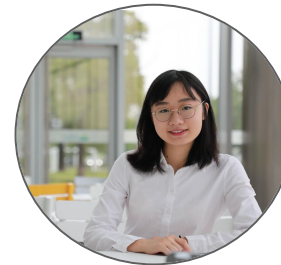
Aditya



Ioana



Farzad



Islina



Nuno



Orquidea



Daniela



Thank you.

