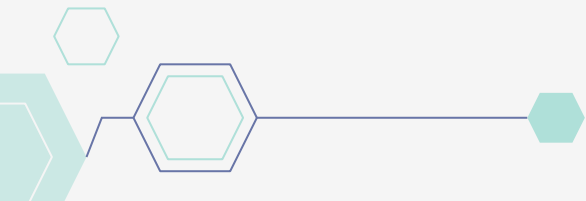




Generating a Land Cover Map for Lebanon

Using Sentinel Hub's eo-learn & GEE's Dynamic World



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01

Introduction






Overview

- Land cover mapping is essential for environmental monitoring, urban planning, and resource management.
- This project focuses on generating a Land Cover Map for Lebanon using:
 - Sentinel Hub's eo-learn (for a single-time land cover classification).
 - Google Earth Engine's Dynamic World (for time-series land cover changes).



Objectives

- Develop a land cover classification map for Lebanon.
 - Implement a time-series analysis to observe land cover changes.
 - Utilize Sentinel-2 satellite data and machine learning techniques for classification.
 - Compare the results from eo-learn and GEE Dynamic World.
- 





Data Sources

- Sentinel-2 Imagery (Optical satellite images with high spatial resolution).
- Sentinel Hub's Land Cover Data (Pre-labeled datasets).
- Dynamic World (Google Earth Engine) (Real-time land cover classification).



02

Tools & Technologies





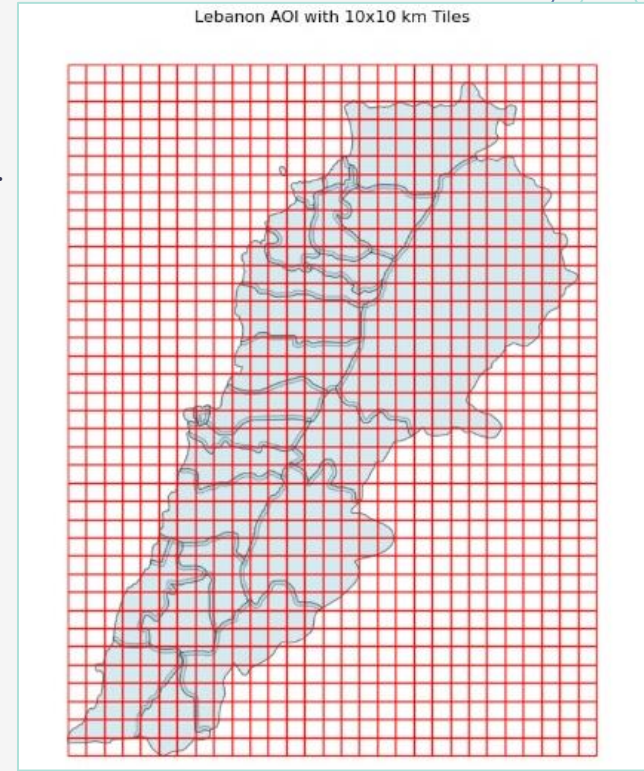
Libraries and Tools

- Python (eo-learn library for data processing).
- Jupyter Notebook (Running the scripts).
- Google Earth Engine (GEE) (For cloud-based analysis).
- QGIS / Matplotlib (For visualization).

Workflow Overview

Approach 1: Using Sentinel Hub's eo-learn

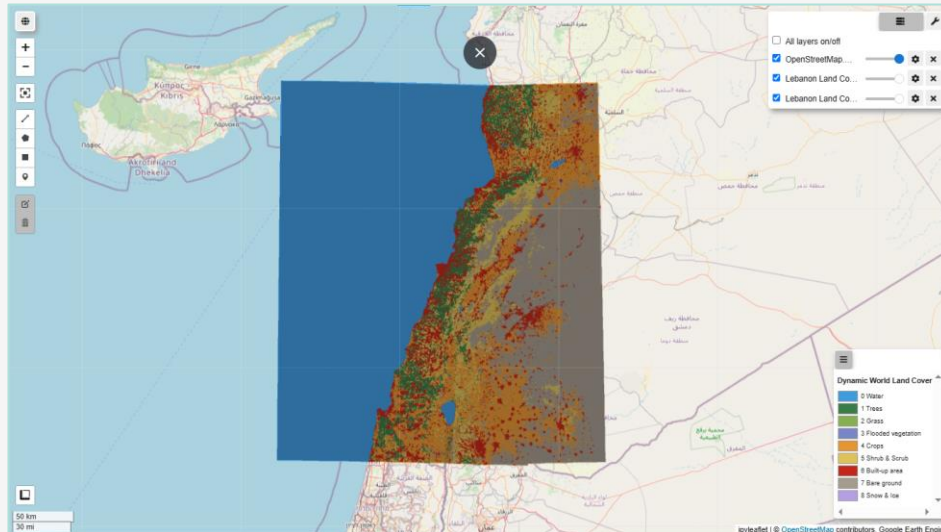
1. Define the Area of Interest (AOI) – Lebanon's boundaries.
2. Data Preparation – Download Sentinel-2 images.
3. Feature Engineering – Compute indices (NDVI, NDWI).
4. Train Machine Learning Model (Random Forest).
5. Classify Land Cover and generate the final map.



Workflow Overview

Approach 2: Using GEE Dynamic World

1. Import Dynamic World dataset into GEE.
2. Select Lebanon's AOI and define a time range.
3. Process the images and classify land cover.
4. Generate a time-series animation of land cover changes.





03

Implementation



Techniques

Sentinel Hub's eo-learn

GitHub Repository:

- SI_LULC_pipeline.ipynb
- Land Cover Map Example

GEE Dynamic World

GEE Notebooks Used:

- Land Cover Time-Series
- Land Cover Classification

Results

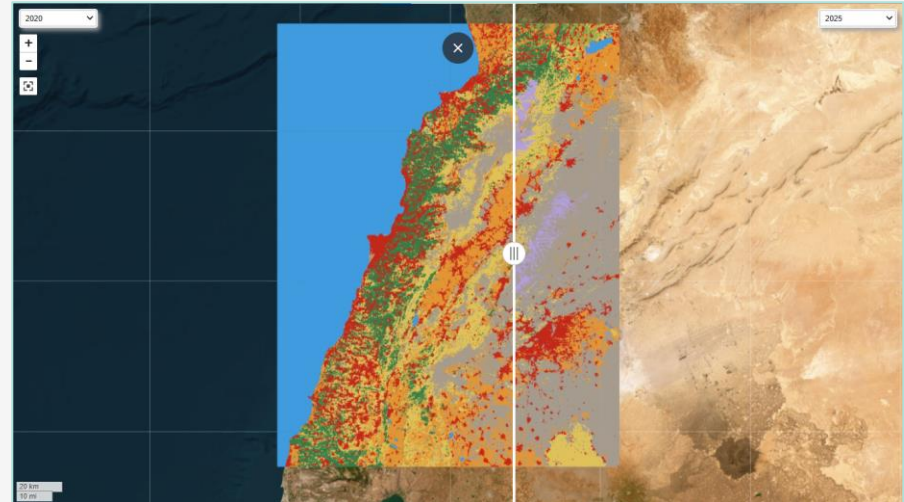
Expected Outcome:

- A classified Land Cover Map identifying urban, forest, water, and agricultural areas
- Insights into land use patterns for better decision-making
- Support for environmental monitoring and urban planning

Results

Time-Series Land Cover Map

- Generated dynamic land cover maps over time.
- Identified land cover changes (e.g., urban expansion, deforestation).
- Time-lapse video shows land cover trends.





04

Challenges



Challenges

- Data Preprocessing: Handling missing and cloudy images.
- Computational Resources: Machine learning training is time-intensive.
- Accuracy: Need for high-quality labeled training data.
- API Limitations: Sentinel Hub and GEE have rate limits.



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Future Improvements



Future Improvements

- Improve classification accuracy with Deep Learning models.
- Integrate multi-temporal analysis for better predictions.
- Use higher-resolution satellite imagery (Sentinel-3, Landsat-9).
- Develop a web-based dashboard for interactive visualization.



06

Conclusion



The slide features decorative chemical structures in the corners. The top-left corner shows a benzene ring with a teal dot at the top and a line extending to the right. The top-right corner contains a complex network of interconnected hexagons and lines, with some teal dots. The bottom-left corner shows a benzene ring connected to a teal hexagon on the right, with another hexagon above it. The bottom-right corner is partially obscured by a teal shape.

Thank You