



# **GEO-AI Challenge for Cropland Mapping**

# Background



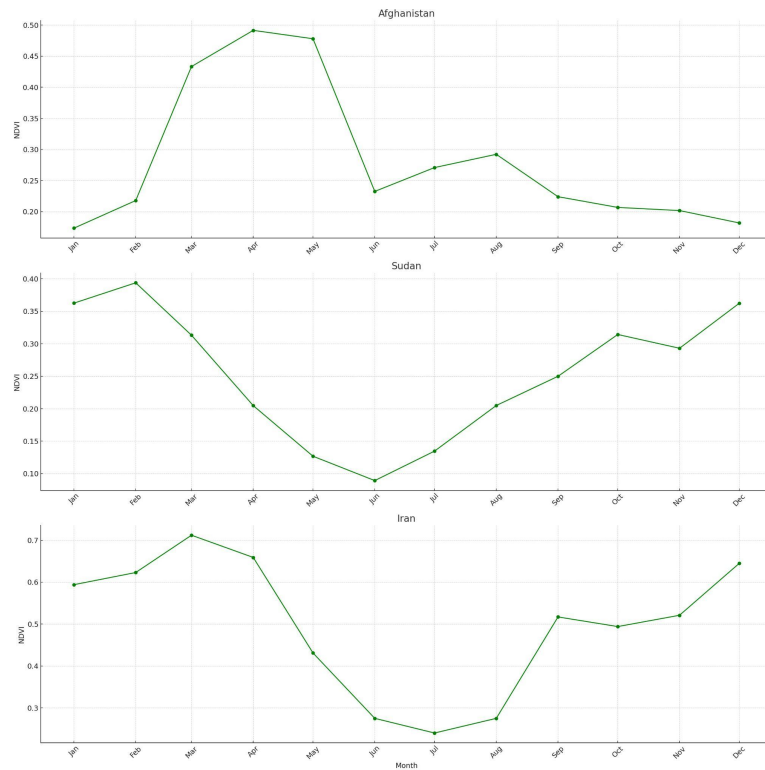
- Timely, accurate crop maps are vital for numerous agricultural applications.
- Limitations of Existing Products: Lack annual updates, reducing usefulness for monitoring changes.
- New Opportunities: High-resolution open data (e.g. Sentinel 2) and machine learning models enable scalable modeling of cropland extent.

# Methodology and WorkFlow



- Crop Season identification using NDVI Temporal Profile
- Sentinel 2 Data Processing - Cloud Masking/Compositing
- Vegetation Indices generation
- Model Development - HyperParameter Tuning and Classification Accuracy
- Crop Extent generation

# Crop Season from NDVI Profile



## Sentinel- 2 Data Processing



- Cloud Masking was done using the Cloud Probability dataset.
- All pixels with a cloud probability of 50 or above were masked out.
- The value of 50 was chosen to ensure a good balance between quality and quantity for the analysis.

# Vegetation Indices

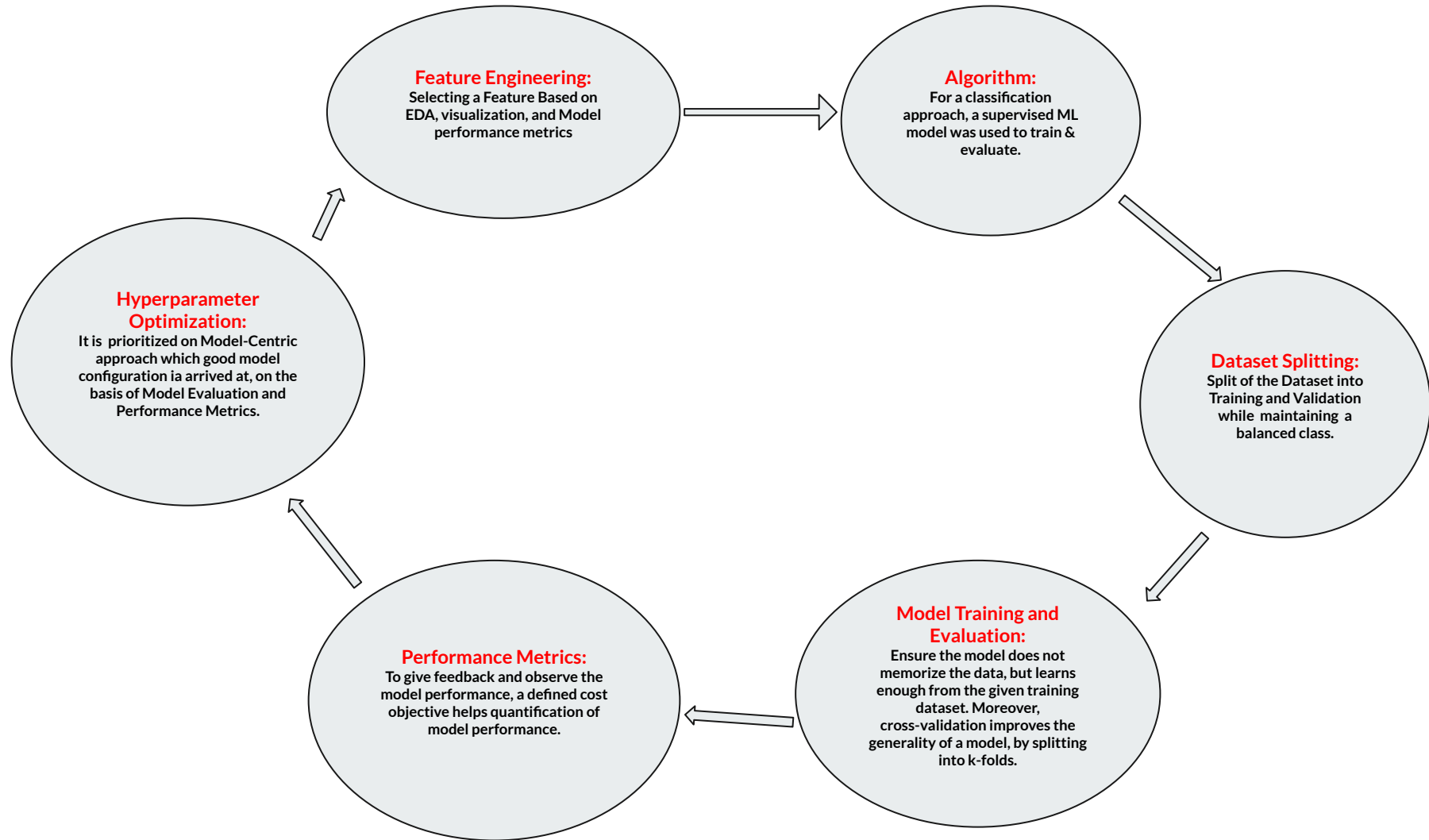


- A wide variety of vegetation Indices were generated for each composite of the crop season to establish contrast between crop and non crop pixels
- Following Indices shortlisted
- NDVI, NDRE, ARVI, RVI, SCI, NDMI, GNDVI, BSI

# Model Development



- It is an iterative process in which many models were derived, evaluated, and tested until achieving the best performance in terms of accuracy results.
- Typically focus on model creation, experimentation, and evaluation.

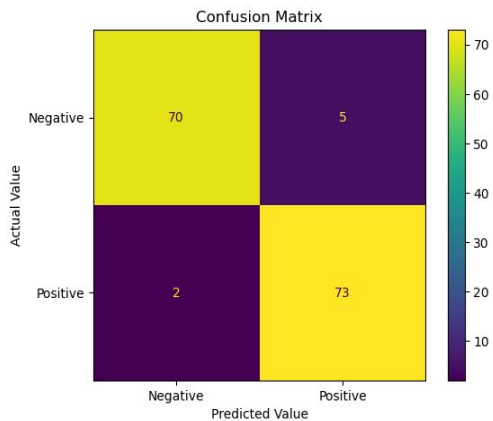




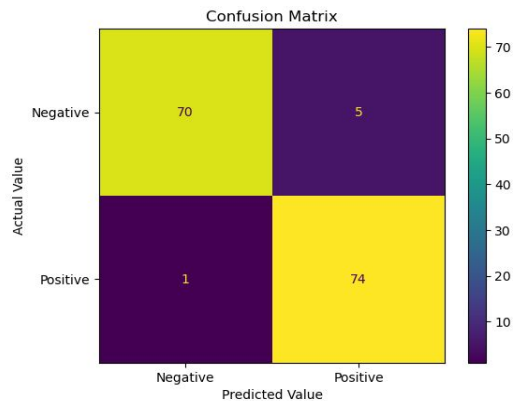
# Classification Accuracy



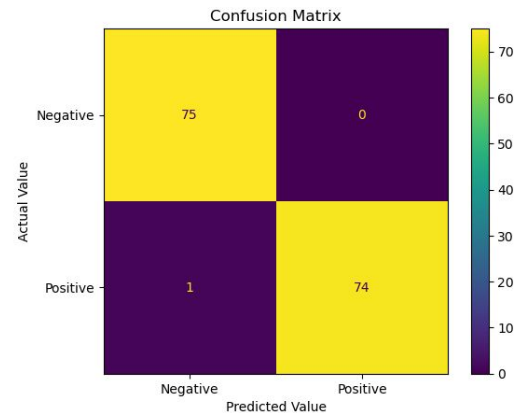
## Afghanistan



## Iran

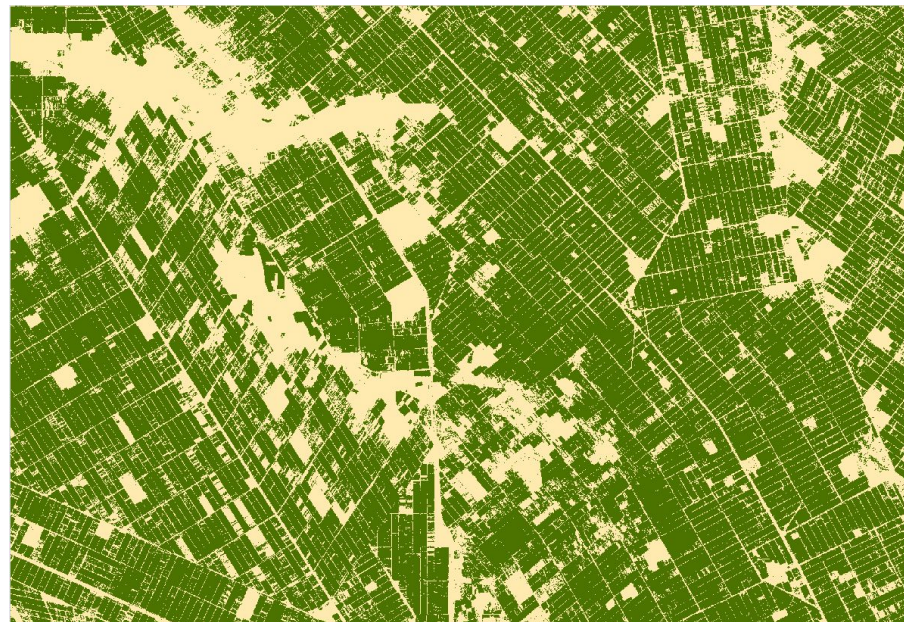
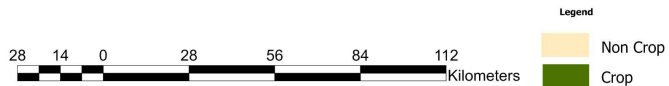
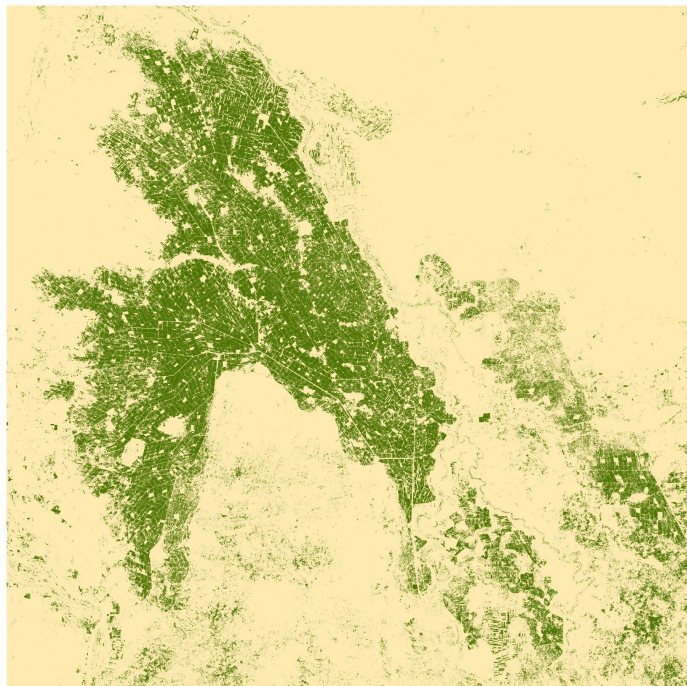


## Sudan

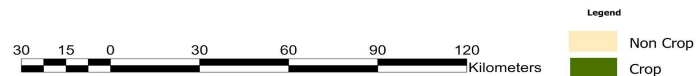
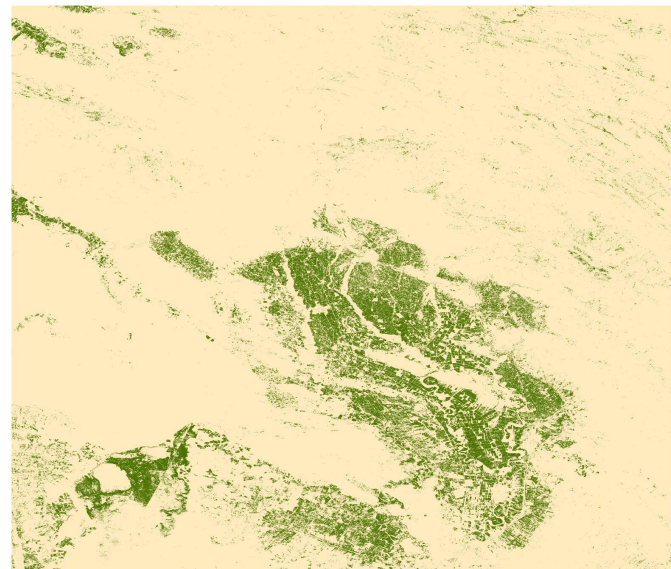


- **Negative: Non-Crop**
- **Positive: Crop**

# Spatial Extent - Sudan



# Spatial Extent - Iran





# Spatial Extent - Afghanistan

