

AI and Remote Sensing for Monitoring Onion under Salinity Stress



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INTRODUCTION

Salinity is a major environmental stressor that adversely affects agricultural productivity globally. In Tunisia, poor water quality is a critical issue, with over 30% of water resources exceeding 4 g/l salinity. Traditional soil sampling methods are often time-consuming and spatially limited, making it challenging to accurately capture soil variability across fields. Remote sensing (RS) offers a rapid, large scale tool to monitor various indices, while artificial intelligence (AI) models can predict soil properties and integrate them with satellite data. **This study combines AI-predicted soil properties with satellite-derived indices to assess onion responses under salinity stress, advancing a data-driven approach for precision agriculture.**



Objective: To monitor onion response to salinity stress using aSpace's AI-driven soil analysis and remote sensing platform

MATERIAL AND METHODS

Split-plot design within RCBD: Main plots: irrigation water type
saline water : 8.40 mS/cm (S-plot) , normal water: 1.83 mS/cm (N-plot)
Sub-plots: onion accessions (analysis focused on overall plot responses).



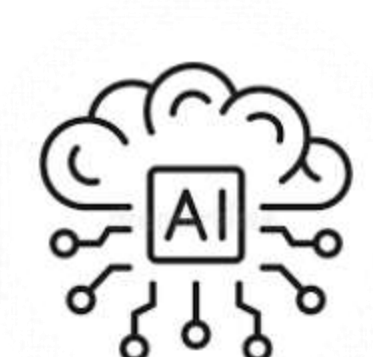
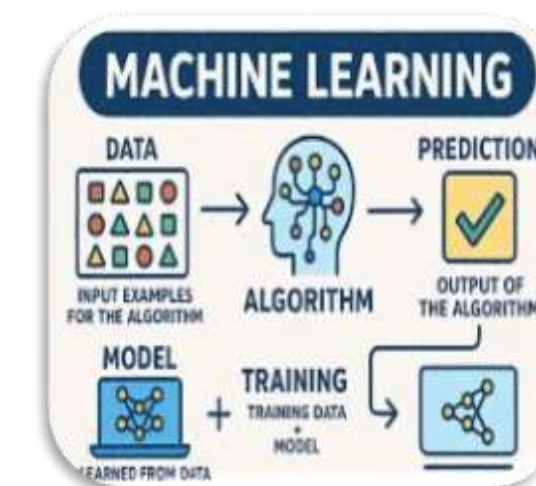
➤ **AI-predicted soil parameters** (organic matter, cation exchange capacity (CEC), electrical conductivity, pH, and N-P-K contents) to characterize soil variability and identify factors influencing onion response



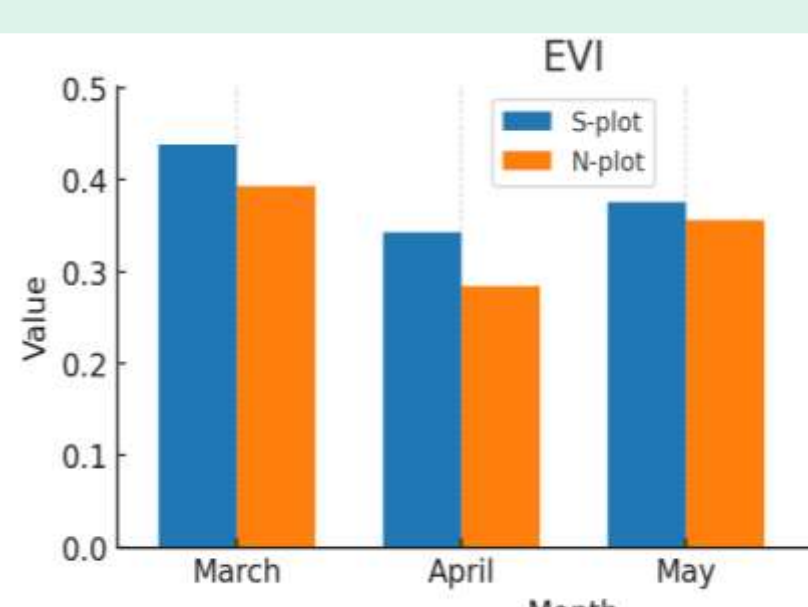
➤ **Multi-spectral, multi-temporal satellite imagery** (Sentinel-2, PROBA-V, Landsat 8, and MODIS)
➤ **Key vegetation and salinity indices** (NDVI, EVI, SAVI, BSI, and NDSI) computed and cluster analysis performed to map healthy vegetation patches within the trial.



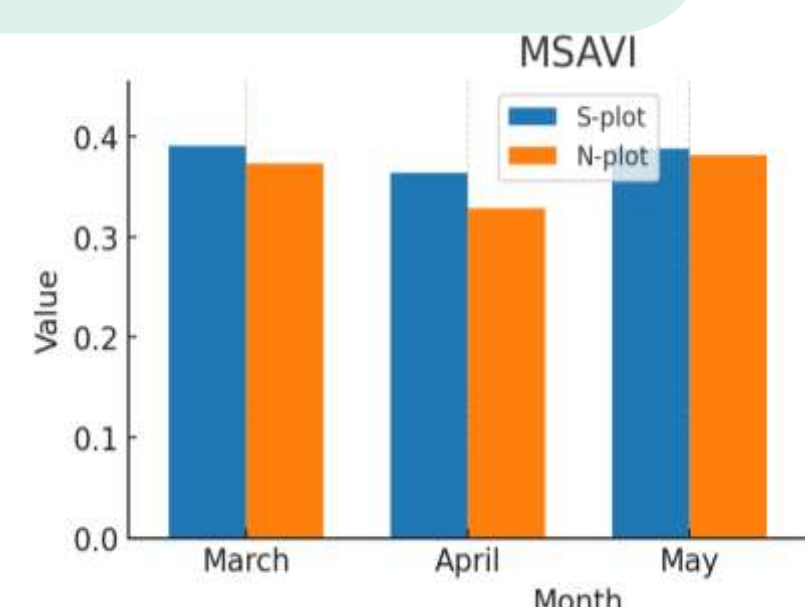
Experimental Station of Sahline



Normalized Difference Vegetation Index



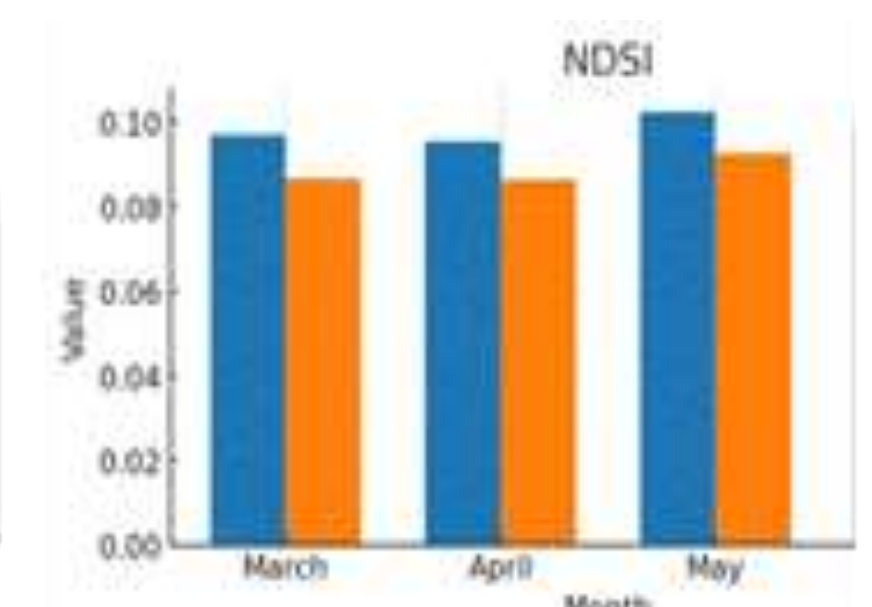
Enhanced Vegetation Index



Soil-Adjusted Vegetation Index



Bare Soil Index



Normalized Difference Salinity Index

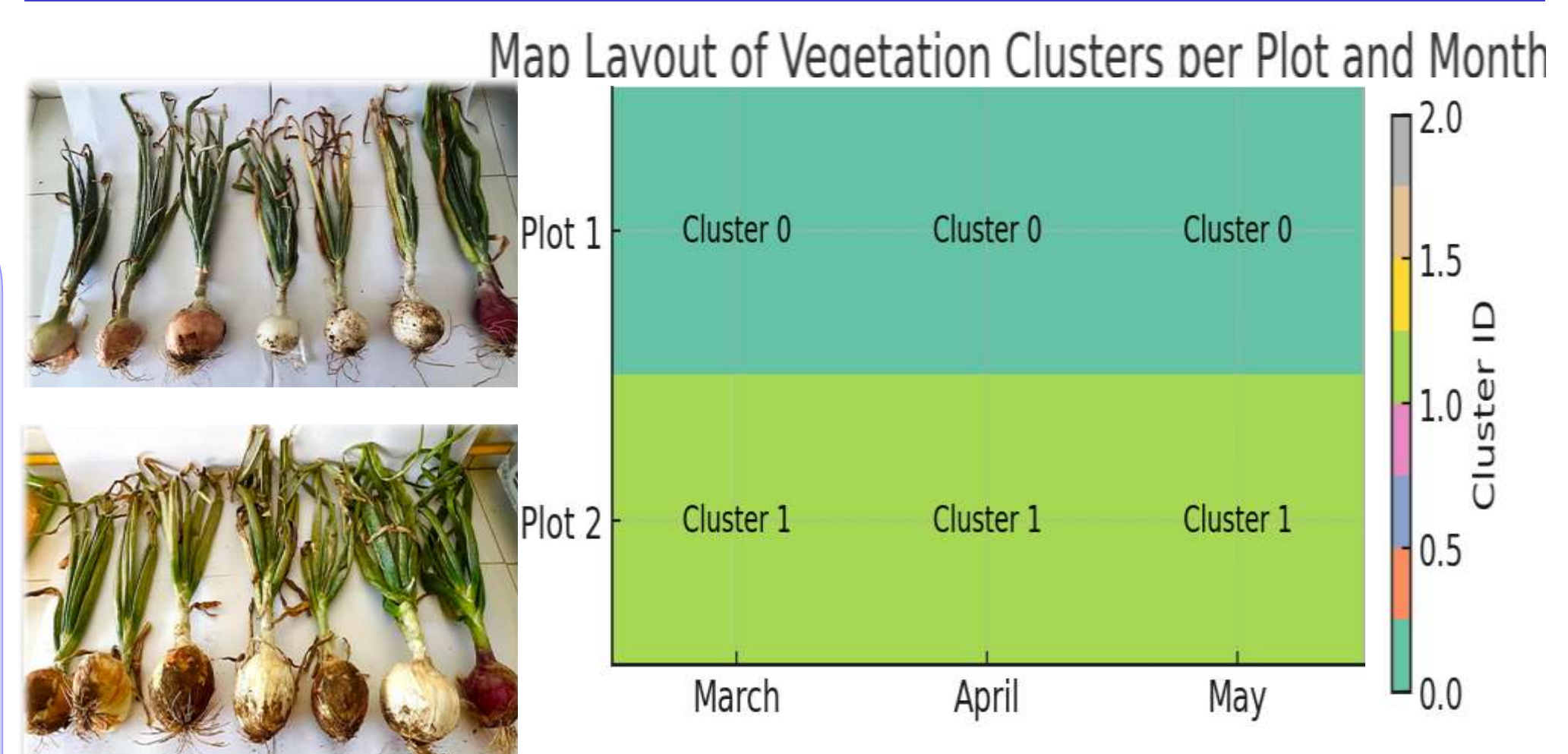
RESULTS

- **NDSI** remained consistently higher in the saline plot throughout the monitoring period, confirming persistent salinity stress in the affected plot and aligning with the lower **NDVI** values observed under saline conditions.
- **BSI** was slightly higher in the non-saline plot, which was unexpected and could be attributed to soil background, moisture effects, or early canopy development differences.
- **MSAVI** and **EVI** were slightly higher in the saline plot across months, possibly due to surface reflectance effects or early-stage physiological responses to stress.

CONCLUSION

- aSpace's AI platform enabled rapid, field-plot-scale estimation of soil properties, overcoming the limitations of traditional sampling methods.
- **NDSI and NDVI** were the most effective indices for detecting salinity effects in onion plots, suggesting their suitability as key indicators in future monitoring and varietal screening efforts.
- This first-year study, demonstrated the potential of AI and remote sensing for rapid, non-destructive assessment of salinity effects. Future trials will refine this approach, enabling more detailed varietal-level evaluation and providing a scalable framework for precise crop selection and assessment of abiotic stress resilience.

AI-Predicted Soil Properties of the Experimental Site		S-plot and N-plot
Soil properties (pH, Carbon content, texture)	pH: 6.7 and carbon content 9 g/kg Sandy clay loam soil : Sand (55%, Clay 26%, Silt 20%)	
Macronutrients (N, P, K, S)	>100 ppm , 6.4 ppm , 80.5 ppm and 12.6 ppm	
Micronutrients (Mg, Ca, Zn)	163 ppm , 991.3 ppm and 1 ppm	
Organic matter	11 g /kg	
CEC (cmol/kg)	8	



Plot 1 (S-plot) shows lower vegetation vigor (Cluster 0) from March to May
Plot 2 (N-plot) displays better vegetation condition (Cluster 1).