



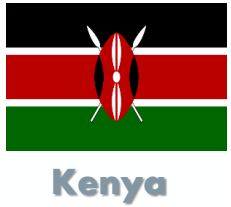
AFRI4CAst Supporting Food Security & Food Safety in Africa

Emanuel Lekakis
Head of Agronomists Dept. AgroApps
mlekakis@agroapps.gr



SAPIENZA
UNIVERSITÀ DI ROMA





Kenya



UNIVERSITY OF NAIROBI

Collected data



- **Agronomic & yield data**
- **LAI field measurements**
- **Mycotoxin production**
- **Rust disease severity**
- **LAI & chlorophyll**

Uganda



Collected data



- **Agronomic & yield data**
- **Rust disease severity**



Environmental Surveys
INFORMATION, PLANNING AND POLICY SYSTEMS








Objectives

- Evaluate the **accuracy of AquaCrop** to predict wheat, maize, rice yield
- Determine **ready-to-use crop files** based on satellite derived phenology
- Provide a **simple, less data-intensive methodology** for future yield prediction

Work Plan

- Locate **wheat, maize and rice** parcels
- Collect agronomic & yield **data**
- Use **databases** on soil & meteorological data
- Acquire **satellite** images
- Characterize growth based on **NDVI** evolution
- Classify **growth patterns**
- **Run** AquaCrop
- Evaluate the **accuracy** of yield prediction

In-situ data

Country	Crop	Location	Parcels	Crop density	Irrigation	Variety	Sowing Harvest	Yield
Kenya		Narok county	36	✓		✓	✓	✓
Kenya		Narok county	74	✓		✓	✓	✓
Kenya		Ahero county	252	✓	✓	✓	✓	✓
Uganda		Bukwo District	149	✓		✓	✓	✓
Uganda		Mubende District	147	✓		✓	✓	✓

EO data

Sentinel - 2

Growing Season



$$NDVI = \frac{NIR - Red}{NIR + Red}$$

$$LAI = 5.7 \cdot GreenWDRVI^2 + 1.7 \cdot GreenWDRVI - 0.08$$

$$GreenWDRVI = \frac{a \cdot NIR - Green}{\alpha \cdot NIR + Green} + \frac{1 - \alpha}{1 + a}$$

$$CC_{RS}(\%) = 94 \cdot [1 - \exp(-0.43 \cdot LAI)]^{0.52}$$

In-situ data

Country	Crop	Location	Parcels	Chlorophyll	LAI
Kenya		Narok county	50	✓	✓
Kenya		Narok county	50	✓	✓

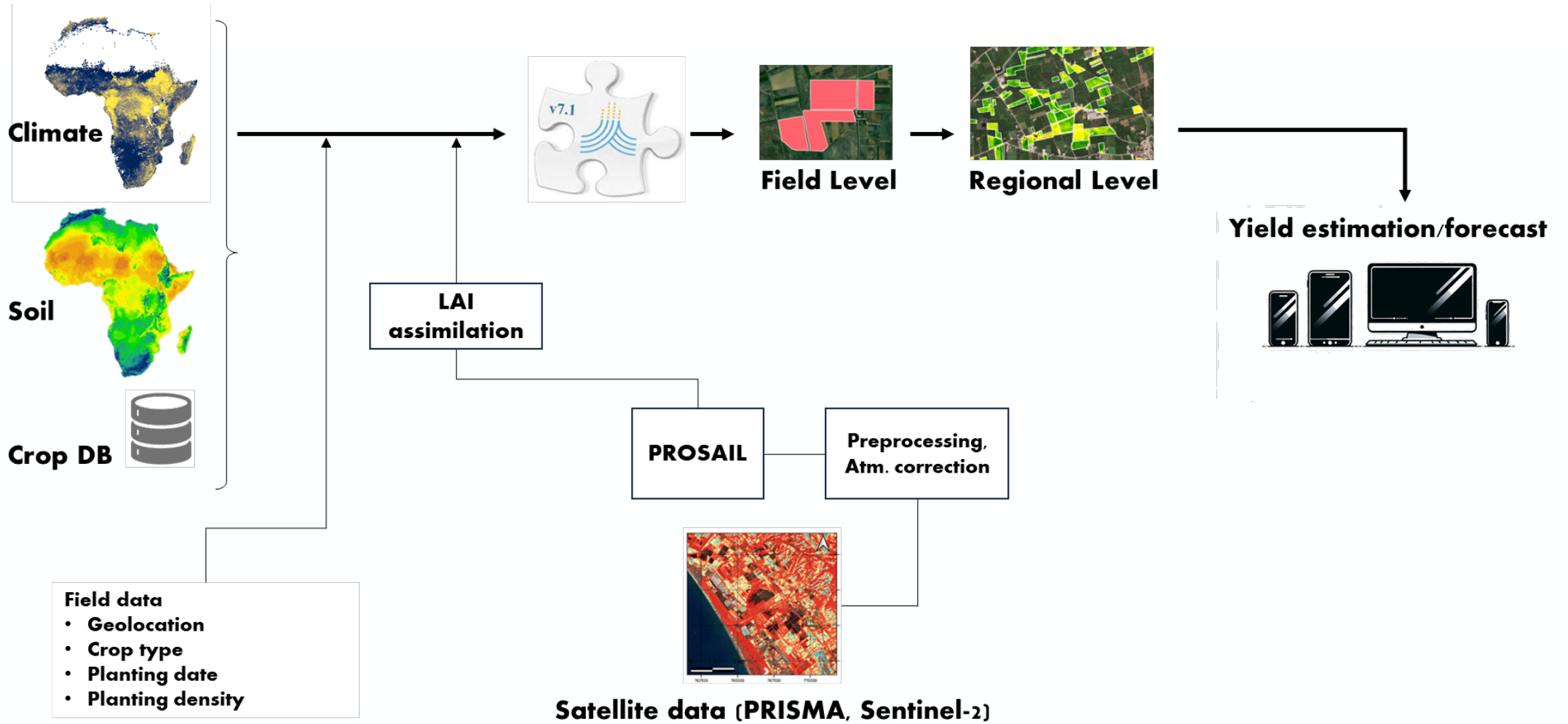


A coupling scheme of PROSAIL with a machine learning regression algorithm to directly extract LAI

EO data



- Spectral channels
- Solar Zenith Angle
- Viewing Zenith Angle
- Relative Azimuth Angle



Objectives

- **Develop or evaluate** a model for rust detection
- **Use satellite data**
- **Provide a simple methodology** for rust detection

Work Plan

- **Locate wheat and maize parcels**
- **Collect in-situ rust detection data**
- **Acquire satellite images**
- **Evaluate the accuracy of rust detection**

A **linear regression** model was developed based on an available cloud-free Landsat-8 image that coincided with the 3rd field visit.

35 LandSat-8 fit indices were tested.




Two approaches:

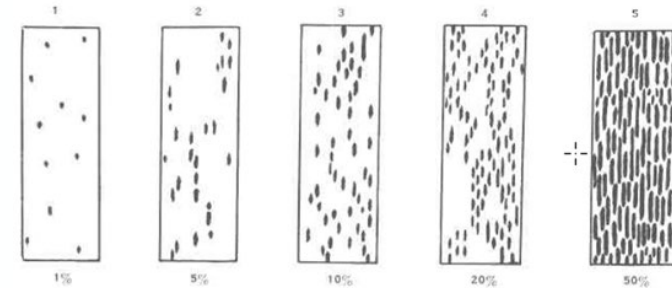
Separate models for each crop in each country.

One model combining all available datasets.

	KENYA		KENYA & UGANDA
Wheat data	Maize data	Wheat & Maize data	Wheat & Maize data
CI	INTENSITY	ALTERATION	WRDVI
WDRVI	WETNESS	WETNESS	BWDRVI
INTENSITY	AFRI ₁₆₀₀	AFRI ₁₆₀₀	variGREEN
AFRI ₁₆₀₀	PVR	FE ₂	GARI
SIWSI		WRDVI	

In-situ data

Country	Crop	Location	Parcels	Rust Scoring	Variety	Sowing Harvest
Kenya		Narok county	50	✓	✓	✓
Kenya		Narok county	50	✓	✓	✓
Uganda		Bukwo District	40	✓	✓	✓



EO data

LandSat-8

Specific dates

Enhanced Vegetation Index 2
 Ferric iron, Fe₂₊
 Ferric iron, Fe₃₊
 Global Vegetation Moisture Index
 Green atmospherically resistant vegetation index
 Green Optimized Soil Adjusted Vegetation Index
 Green Soil Adjusted Vegetation Index
 Green-Blue NDVI
 Green-Red NDVI
 Ideal vegetation index
 Intensity

Aerosol free vegetation index 1600
 Alteration
 Ashburn Vegetation Index
 Atmospherically Resistant Vegetation Index
 Atmospherically Resistant Vegetation Index 2
 Blue-wide dynamic range vegetation index
 Chlorophyll Index Green
 Coloration Index
 Green Normalized Difference Vegetation Index
 Differenced Vegetation Index MSS
 Enhanced Vegetation Index

Mid-infrared vegetation index
 Norm G
 Norm NIR
 Norm R
 Normalized Difference Plant pigment ratio
 Normalized Difference Photosynthetic vigor ratio
 Normalized Difference 860/1640
 Specific Leaf Area Vegetation Index
 Tasseled Cap – vegetation
 Tasseled Cap – wetness
 Weighted Difference Vegetation Index
 Wide Dynamic Range Vegetation Index

Objectives

- **Develop or evaluate** a model for aflatoxin B₁ and deoxynivalenol (DON) prediction
- Use **agro-meteorological data**
- Provide a **simple methodology** for aflatoxin contamination risk prediction

Work Plan

- Locate **maize** parcels
- Perform maize kernels **sampling & analysis**
- Acquire **meteorological data**
- Evaluate the **accuracy** of the mechanistic weather-driven models

Risk Levels	Aflatoxin	Deoxynivalenol
no risk	9.6% (n=5)	-
very low risk	-	-
low risk	90.4% (n=47)	78.8% (n=41)
moderate risk	-	7.7% (n=4)
high risk	-	3.8% (n=2)
very high risk	-	9.6% (n=5)

In-situ data

Country	Crop	Location	Parcels	Aflatoxin DON	Aflatoxin AFB ₁	Variety	Sowing Harvest
Kenya		Narok county	52	✓	✓	✓	✓



Meteorological data & tested models

Model Input:

- Temperature (min, max, average)
- Relative Humidity
- Cumulative daily Precipitation

ERA-5, ERA-5 Land

Aflatoxin model

Aflatoxin B₁ Index: AFI

AFI < 40 = no risk

40 < AFI ≤ 120 = low risk

120 < AFI ≤ 150 = intermediate risk

150 < AFI ≤ 180 = high risk

Deoxynivalenol model

Five levels of risk for DON

DON ≤ 0.1 indicates very low risk

0.1 < DON ≤ 0.9 low risk

1 < DON ≤ 1.9 intermediate risk

2 < DON ≤ 4,9 high risk

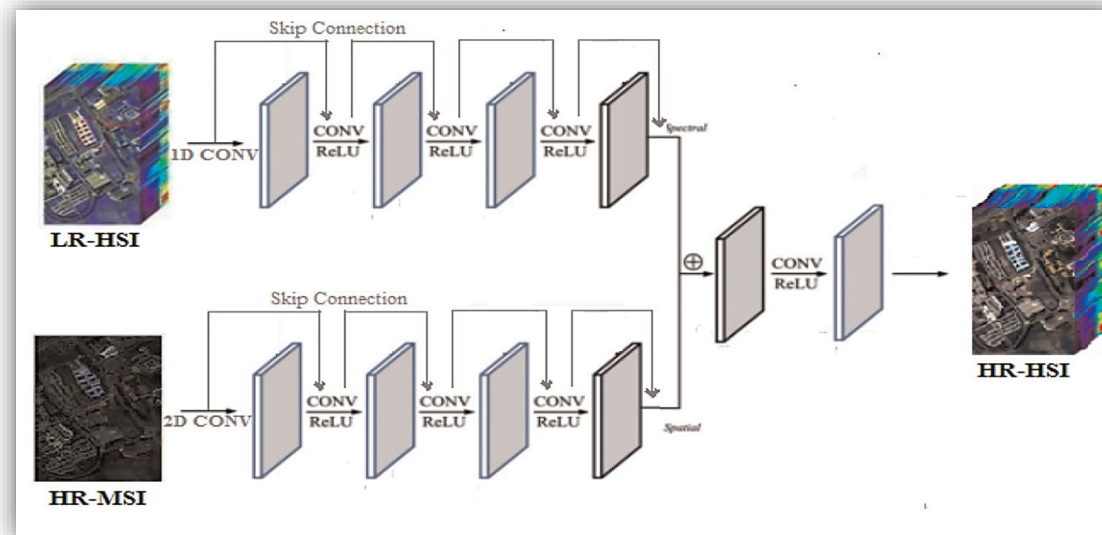
DON > 5 very high risk

Objectives

- **Spatial Enhancement** of hyperspectral imagery
- **Downscale from 30 m to 10 m** spatial resolution
- Apply a **Deep Learning** model: **Guided Deep Decoder (GDD)** for unsupervised HS-MS fusion & **Zero-Shot Learning**

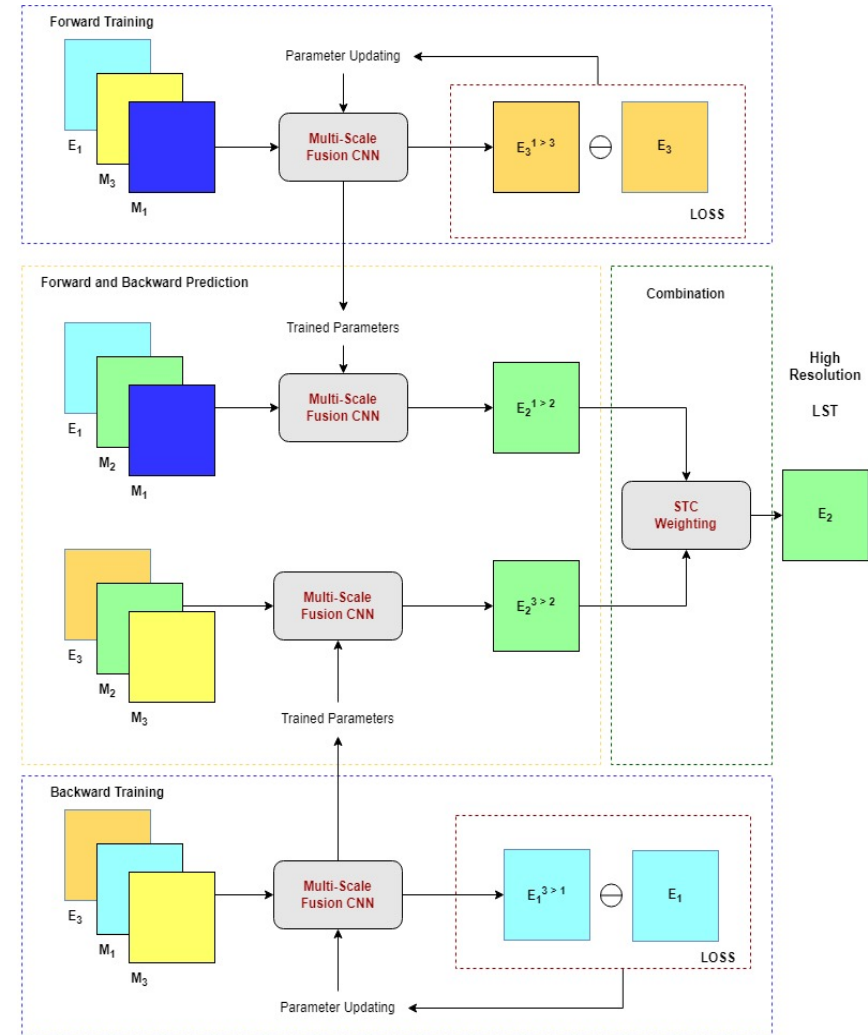
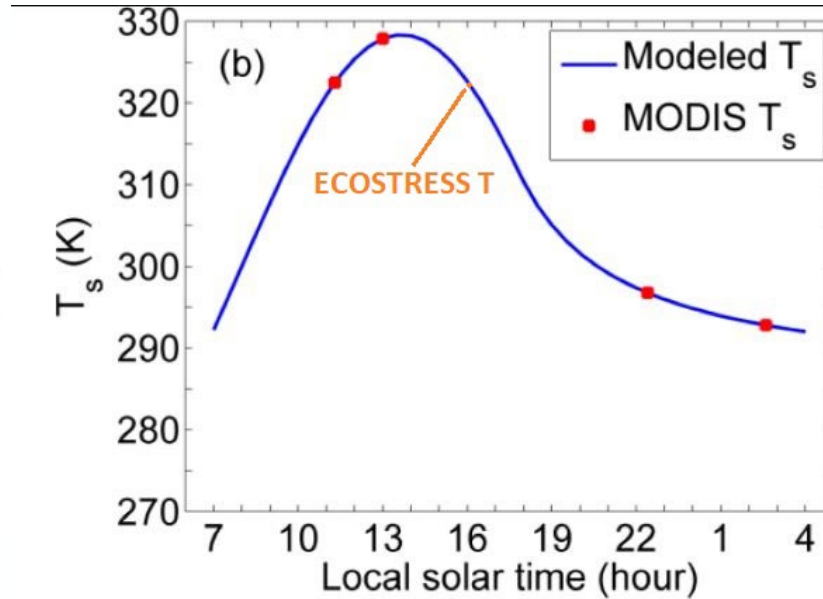
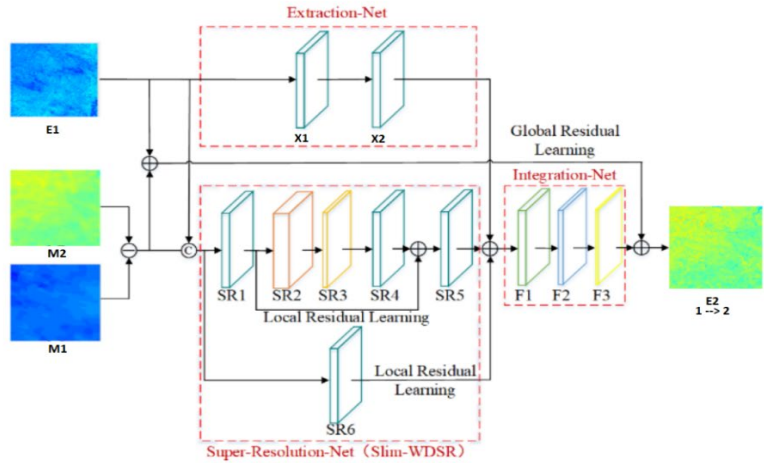
Work Plan

- **Locate PRISMA & S2** images depicting the same **area**
- **Acquire the cloud free satellite images**
- **Evaluate** the DL model



Objectives

- **Diurnal Temperature Cycle (DTC) modeling with MODIS LST data**
- **External validation of DTC with Sentinel-3 LST**
- **Spatiotemporal Fusion Network**





AFRI4CAst Supporting Food Security & Food Safety in Africa

Emanuel Lekakis
Head of Agronomists Dept. AgroApps
mlekakis@agroapps.gr



SAPIENZA
UNIVERSITÀ DI ROMA

