# Using Earth Observation in Crop Yield Modeling: Latest Development

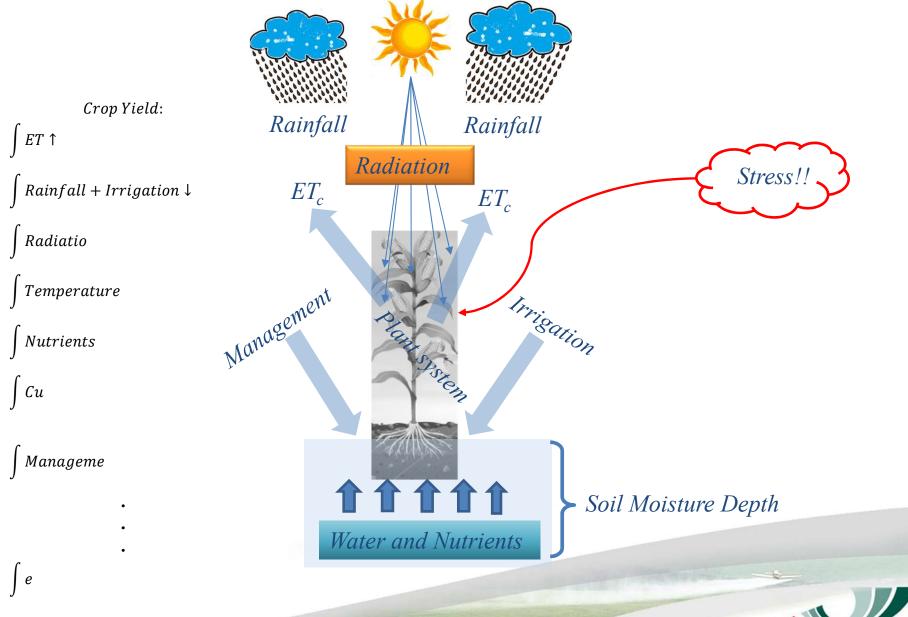
Part(1): EO in Crop models: Mechanistic and/or Empirical

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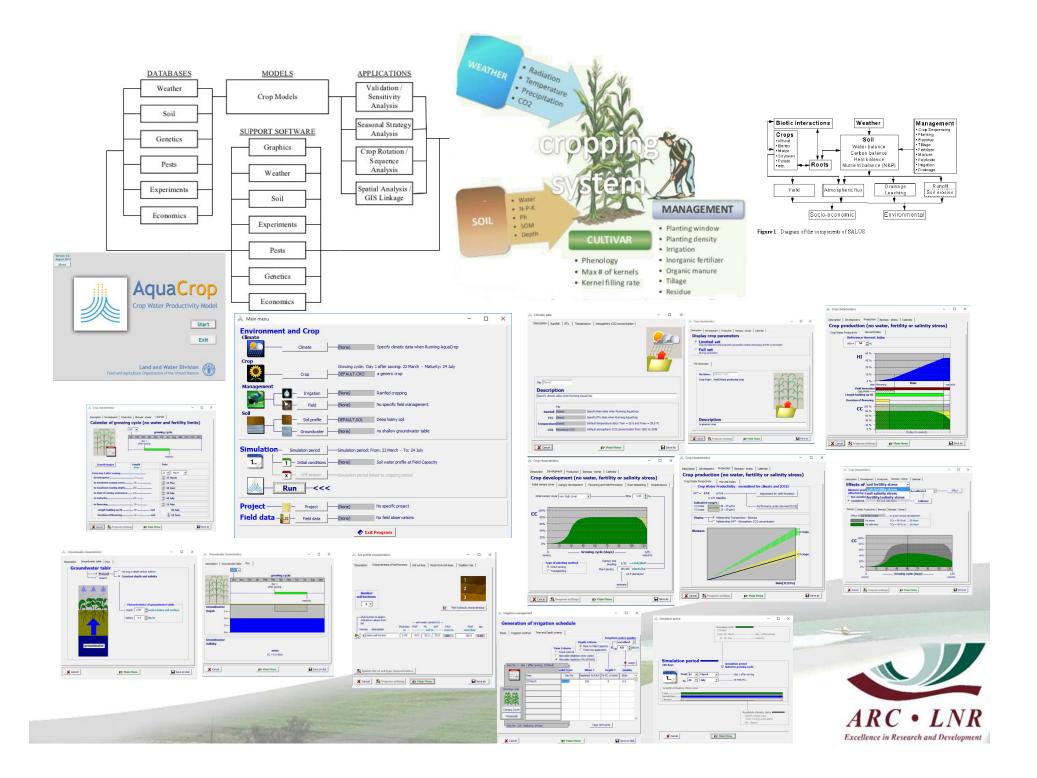


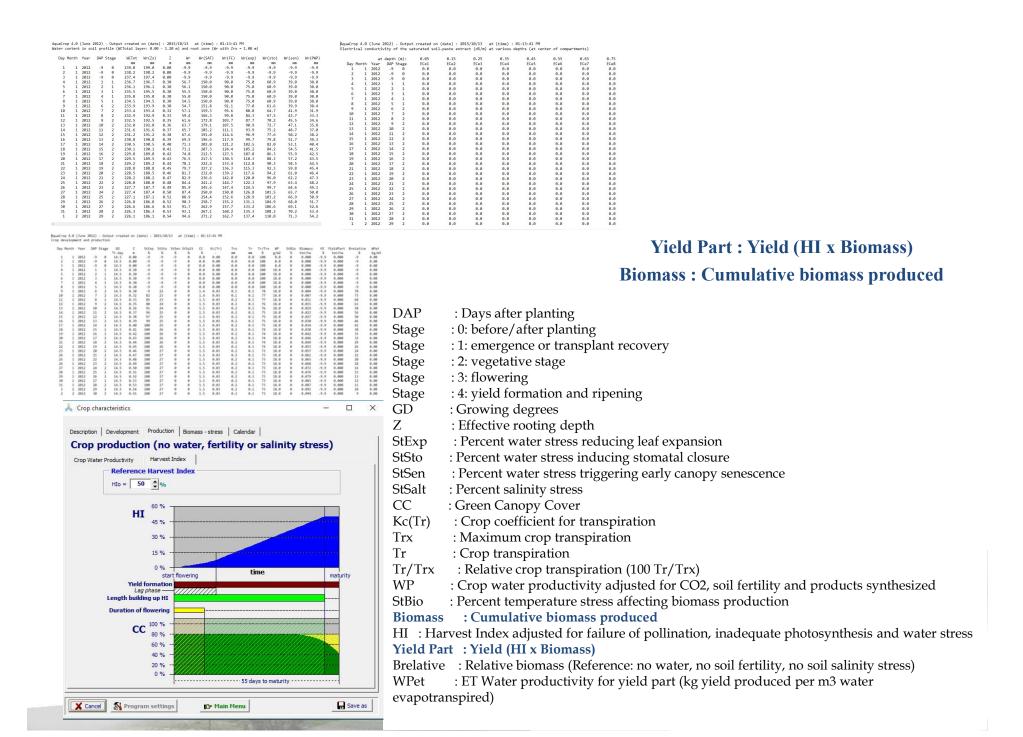
- → Plant system and yield complexity
- →Yield estimation and EO systems
  - → Empirical models
  - → **M**echanistic models
  - **→**Combined models
- → Models Classification
- **→EO** sensors for plant monitoring

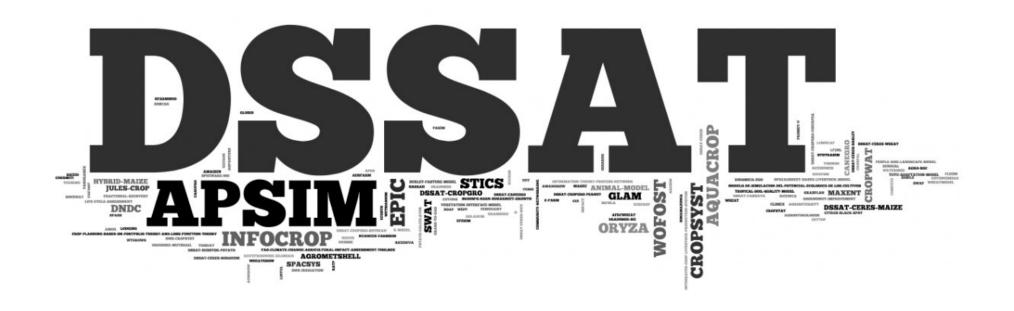




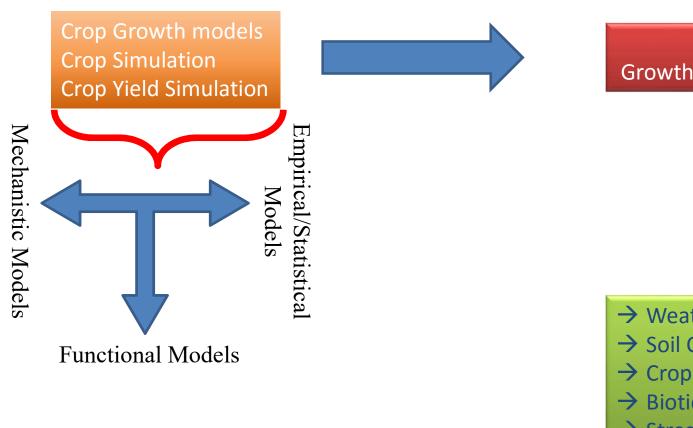




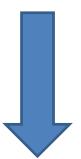








# Crop Growth/Production



- → Weather Conditions
- → Soil Conditions
- → Crop Management
- → Biotic Factors
- → Stress



# **Empirical/Statistical Models**

Empirical models are simply a regression analysis between factor(s) that affect the plant growth and crop yield

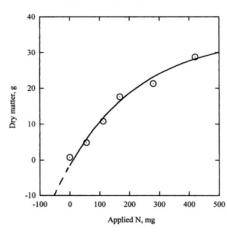
- ➤ Linear Regression
- > Growth Indices
- > Principal Component
- ➤ Markov Chain

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + e$$

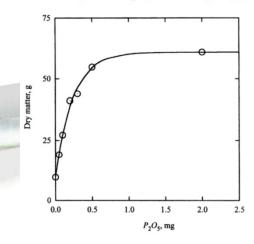
$$Y = \beta_0 + \sum \beta_i G_i + e \qquad G_i = \sum_{w=n_1}^{n_2} r_{iw} X_{iw}$$

$$Y = Y_0 + (Y_m - Y_0)[1 - \exp(-cN)]$$

$$Y_m - Y = (Y_m - Y_0) \exp(-cN)$$



$$Y = 9.6 + 51.4[1 - \exp(-4.16P_2O_5)]$$

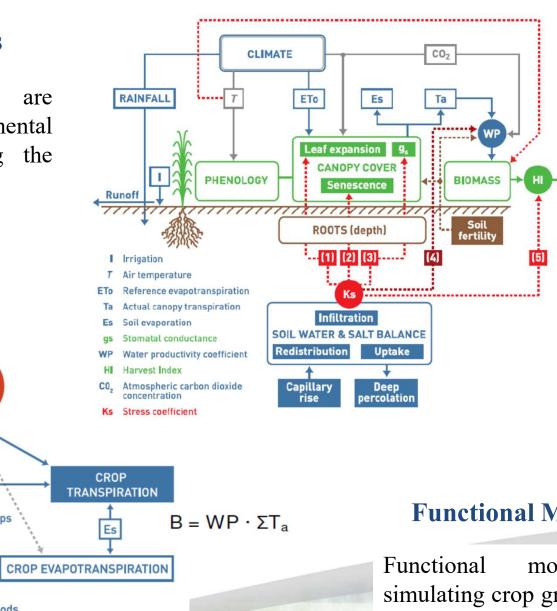


## **Mechanistic Models**

Mechanistic models are simulating the fundamental mechanisms governing the crop growth and yield

**BIOMASS** 

YIELD



#### long-term periods

daily time steps

SOLAR

RADIATION

$$\left(\frac{\mathbf{Y}_{x} - \mathbf{Y}_{a}}{\mathbf{Y}_{x}}\right) = k_{y} \left(\frac{\mathbf{ET}_{x} - \mathbf{ET}_{a}}{\mathbf{ET}_{x}}\right)$$

#### **Functional Models**

YIELD

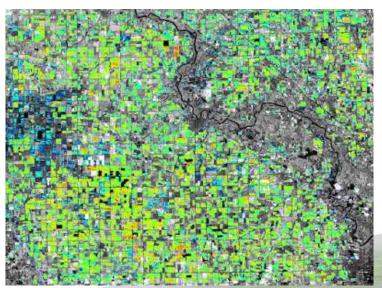
models are simulating crop growth using graphical interpretation of the fundamental processes (System Dynamics)

## Earth observation potential in crop yield estimation?

Several satellites has a capability to measure the land reflectance at spectral range that can be used for plant monitoring

Major ranges for plant monitoring are "Red" and "Infrared"

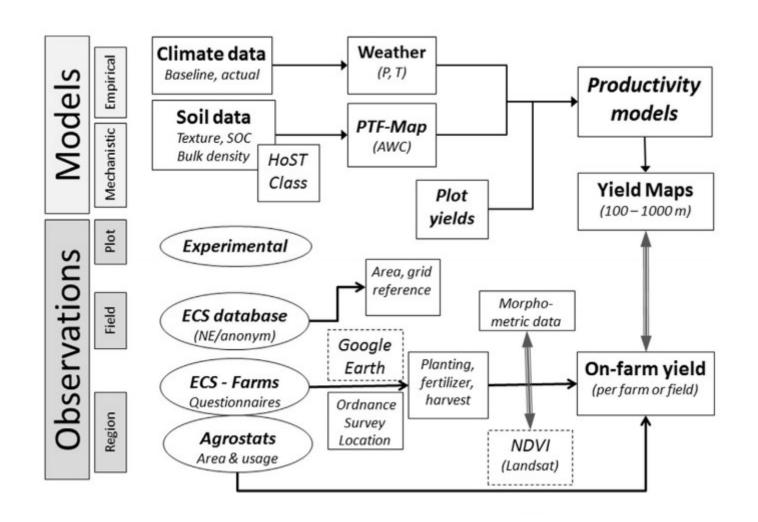
However, the ranges in the shortwave infrared and Thermal inferred (TIR) are also valuable for developing several plant growth indicators related to moisture content in plant and soil.



Scalable Satellite-based Crop Yield Mapper

Based on statistical relationship between NDVI and ground data at 29,000 samples. Model run under Google's Earth Engine





(*Richter et al., 2016*)



Basic parameters that has impact on crop growth and can be monitored by satellite are including:

#### Growth:

- → Biomass production
- → Leaf area
- → Vegetation indices and chlorophyll content

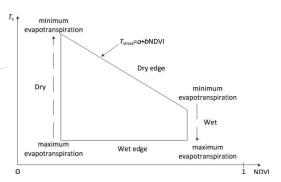
#### Water:

- → Evapotranspiration
- → Water deficiency (Dryness index)
- → Water Use Efficiency

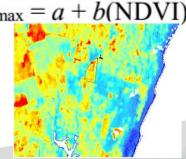
#### **Nutrients:**

 $\rightarrow$  Leaf N (kg/ha)

Indices*	Formulae*	References
NDVI	NIR-Red NIR+Red	Rouse et al. (1974)
EVI	$G\frac{NIR-Red}{NIR+C_1Red-C_2Blue+L}(1+L)$	Huete (1988)
Mid-infrared index	MIR SWIR	Musick and Pelletier (1988)
MSI	MIR NIR	Rock et al. (1986)
NDVI green	NIR-Green NIR+Green	Gitelson et al. (1996)
NDWI	NIR – SWIR NIR + SWIR	Gao (1996)
RSR	$\frac{NIR}{Red} \left( 1 - \frac{SWIR - SWIR_{min}}{SWIR_{max} - SWIR_{min}} \right)$	Brown et al. (2000), Chen et al. (2002)
SR	NIR Red	Birth and McVey (1968), Chen et al. (2002)
TVI	$\left(\frac{\text{NIR-Red}}{\text{NIR+Red}} + 0.5\right)^{1/2} \times 100$	Nellis and Briggs (1992)



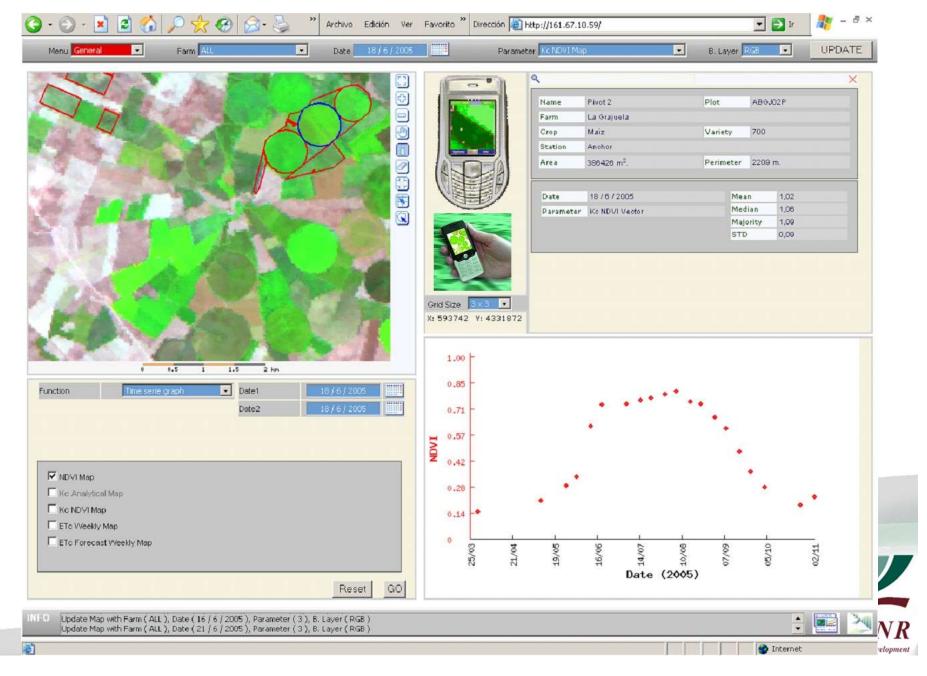
TVDI = 
$$(T_s - T_{smin})/(T_{smax} - T_{smin})$$
  
 $T_{smax} = a + b(NDVI)$ 

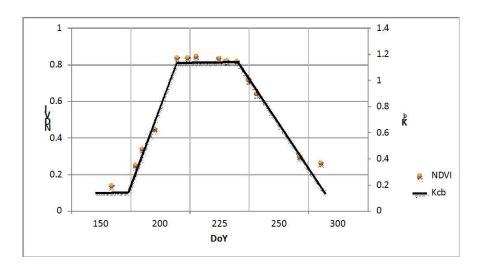


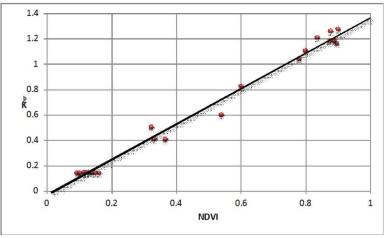
$$NDNI = \frac{\log\left(\frac{1}{\rho_{1510}}\right) - \log\left(\frac{1}{\rho_{1680}}\right)}{\log\left(\frac{1}{\rho_{1510}}\right) + \log\left(\frac{1}{\rho_{1680}}\right)}$$

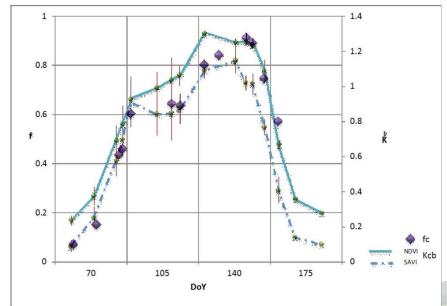


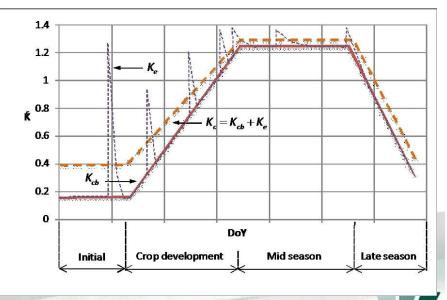
# (Rocha et al., 2012)



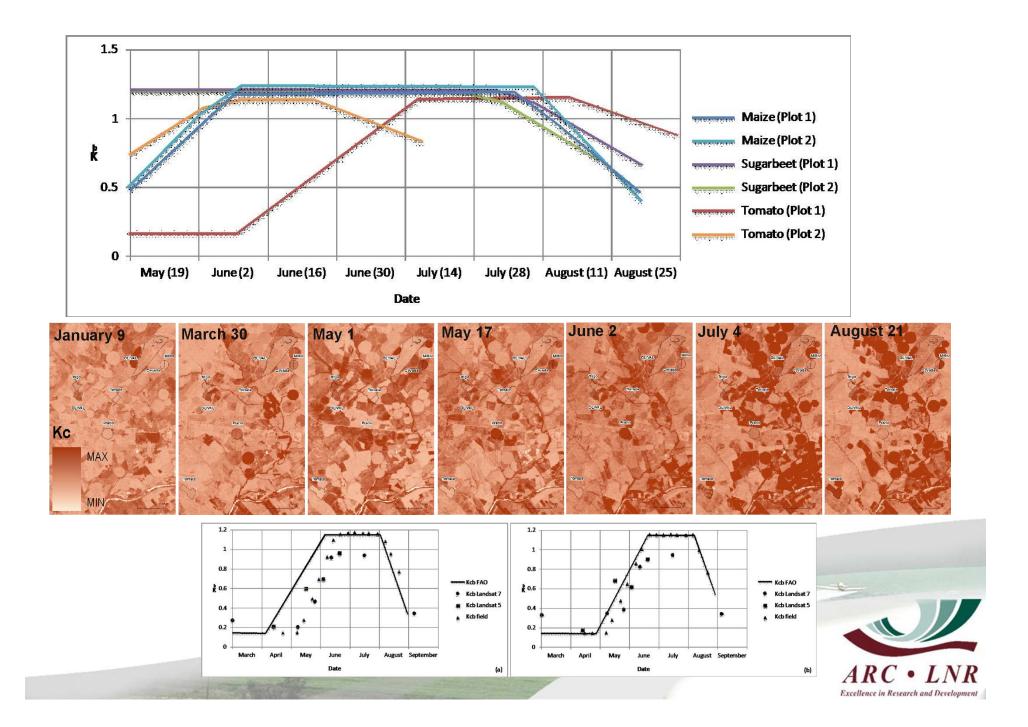


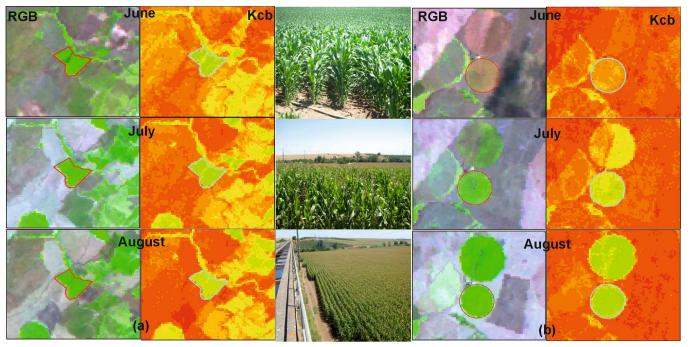


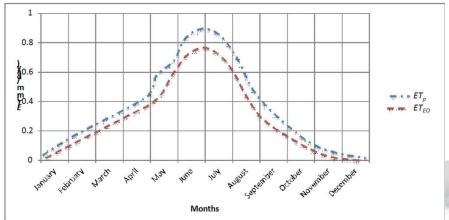




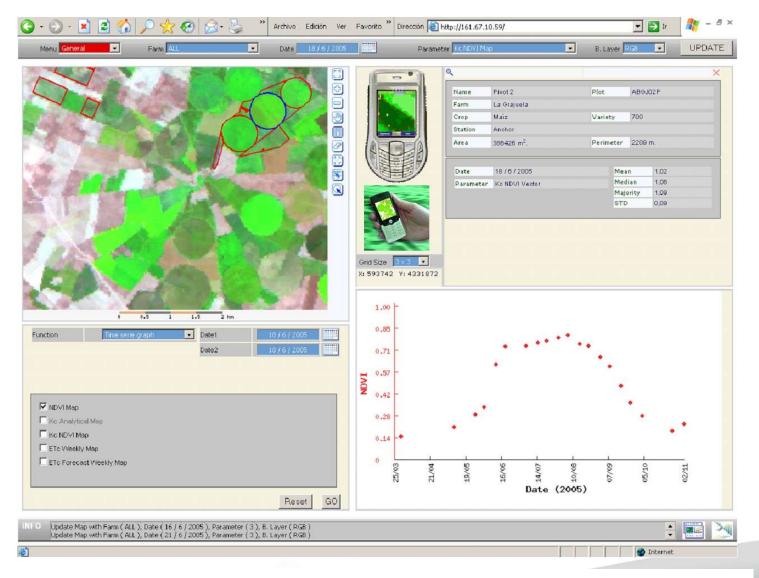












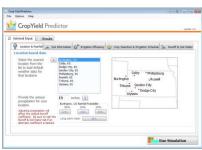
## **PLEIADES**

(Participatory multi-Level EO-assisted tools for Irrigation water management and Agricultural Decision-Support)















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