

Characterization of cereals yields using different models based on remote

sensing indicators from the Sentinel-1 and Sentinel-2 images in a semi-

arid region

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Problematic and context



Source: Office des Céréales https://www.oc.com.tn/



- For Tunisia, net cereal imports between 2005 -2021 amounted on average to 72% of annual consumption.
- An operational early grain yield prediction system has been needed to assist policy makers in making initial assessments and planning for annual grain imports.





Different methodologies have been proposed over the last decades, for the early grain yield prediction using specific multi- and hyper-spectral sensors.

The main objective of this study is to develop empirical models for early prediction of cereal yields at a provincial scale :

• Evaluating the performance of combining several vegetation indices to improve yield prediction results.

- Developing early forecasting models using multisource data as predictors (Sentinel-2 and Sentinel-1 images), and experimental measurements made on cereal test plots over two agriculture years (2017-2018 and 2020-2021): → Regression model
 - \rightarrow Machine learning algorithms

Data base



A total of 82 farms were sampled, which covered nearly 280 ha.

* Khlif, M.; Escorihuela, M.J.; Chahbi Bellakanji, A.; Paolini, G.; Kassouk, Z.; Lili Chabaane, Z. Multi-Year Cereal Crop Classification Model in a Semi-Arid Region Using Sentinel-2 and Landsat 7–8 Data. Agriculture 2023, 13, 1633. https://doi.org/10.3390/agriculture13081633

Methods



Index		Index name	Formula		
NDVI		Normalized Difference Vegetation Index	$\frac{\text{NIR} - \text{RED}}{\text{NIR} + \text{RED}}$		
GNDVI		Green Normalized Difference Vegetation Index	$\frac{\text{NIR} - \text{GREEN}}{\text{NIR} + \text{GREEN}}$		
EVI		Enhanced Vegetation Index	$2.5 * \frac{(PIR - RED)}{(PIR + 6 * RED - 7.5 * BLUE + 1)}$		
EVI2		Enhanced Vegetation index 2 bands	$2.5 * \frac{(PIR - RED)}{(PIR + 6 * RED - 7.5 * BLUE + 1)}$		
NGRDI		Normalized Green Red Difference Index	$\frac{\text{GREEN} - \text{RED}}{\text{GREEN} + \text{RED}}$		
NDRE		Normalized Difference Red Edge	NIR – REDedge NMR + REDedgæ		
CCCI		Canopy Chlorophyll Content Index	$\frac{\text{NIR} + \text{REDedge}}{\text{NIR} - \text{RED}}$ $\frac{\text{NIR} + \text{RED}}{\text{NIR} + \text{RED}}$		
MCARI		Modified Chlorophyll Absorption in Reflectance Index	((VNIR – RED) – 0.2 * (VNIR – GREEN)) * (VNIR/RED)		
DVI		Difference Vegetation Index	NIR – RED		
T۱	/I	Triangular Vegetation Index	0.5 * (120 * (NIR – GREEN) – 200 * (RED – GREEN))		
	bacl	kscatter coefficient VH	VH		
	bac	kscatter coefficient VV	VV		
	tł	ne polarisation ratio	VH/VV ⁵		

The first approach : Regression model



Evolution of vegetation indexes during the two agricultural years 2017-2018 and 2020-2021.

Correlation between different indexes and cereal yield (grain and straw) for the two agricultural years 2017-2018 and 2020-2021.

The first approach : Regression model



Evolution of the NDVI and the VH/VV ratio during the 2017-2018 agricultural year for a plot of barley The VH/VV ratio is sensitive to the growth cycle of the culture and depends on the phenology stade of the plant

The possibility of early estimation of cereal yields using Radar images

The first approach : Regression model



The behaviors of NDVI, VH/VV ratio, VH and VV on wheat field are like barley field from seeding to harvest. → Similar structure and functioning



Evolution of grain yield for 12 March 2018



Evolution of grain yield for 21 March 2021

For the two agriculture years different cultural practices was applied : In 2021 intensive cultivation

The second approach : Machine Learning Algorithms

Approaches based on the most used machine learning algorithms : SVM, RF, Neurons, and XGBoost.



Red-Edge index : GNDVI; NDRE ; CCCI Statistical metrics for the forecasting models in March (3 month before harvest).

Machine Learning	Grain yield		Straw yield	
Algorithm	R ²	RMSE	R ²	RMSE
ANN	0.87	6.08	0.71	10.27
XGBoost	0.89	5.75	0.75	9.6
SVM	0.89	5.74	0.77	9.21
RF	0.89	5.58	0.72	10.08

→ The combination of GNDVI; NDRE ;CCCI performed better than models based on other combination.

→ Random Forest algorithm is the best tool for estimating seed and straw yields.



The second approach : Machine Learning Algorithms

Spatialisation of the grain and straw yields for each cereal plots on the plain of Kairouan

Random Forest

algorithm





9°52'30"

9°52'30"E

I'M ZONE 32N

10°3'0"E

9°42'0"E

Grain yield RF 2020/2021

 $R^2 = 89\%$

RMSE = 5.5

9°42'0"E



The agriculture year (2017-2018)



The agriculture year (2020-2021)

- **Regression Model :** Strong correlation between measured and estimated yield by Rededge indices from mid-March ($R^2_{grain} = 0,78$; $R^2_{straw} = 076$).

- Radar time series allow to capture short phenological stages thanks to the high sensitivity of polarization ratio to plant phenology.

-The change in farming practices in the plain has influenced the capacity to predict yield using the polarization ratio

- Machine Learning Algorithms : Random Forest algorithm is the best tool for early estimating seed and straw yields in March (3 month before harvest).

- Red-Edge indices have demonstrated their performance for early estimation of cereal yields



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Thank you for your attention !









