CGMS-Maroc: National System for Agrometeorological monitoring

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- Part1: CGMS-Maroc
  - Objectives of the system
  - Presentation of the system
- Part 2: Operational cereal yield forecasting in Morocco.
  - Material and methods
  - Results and discussion

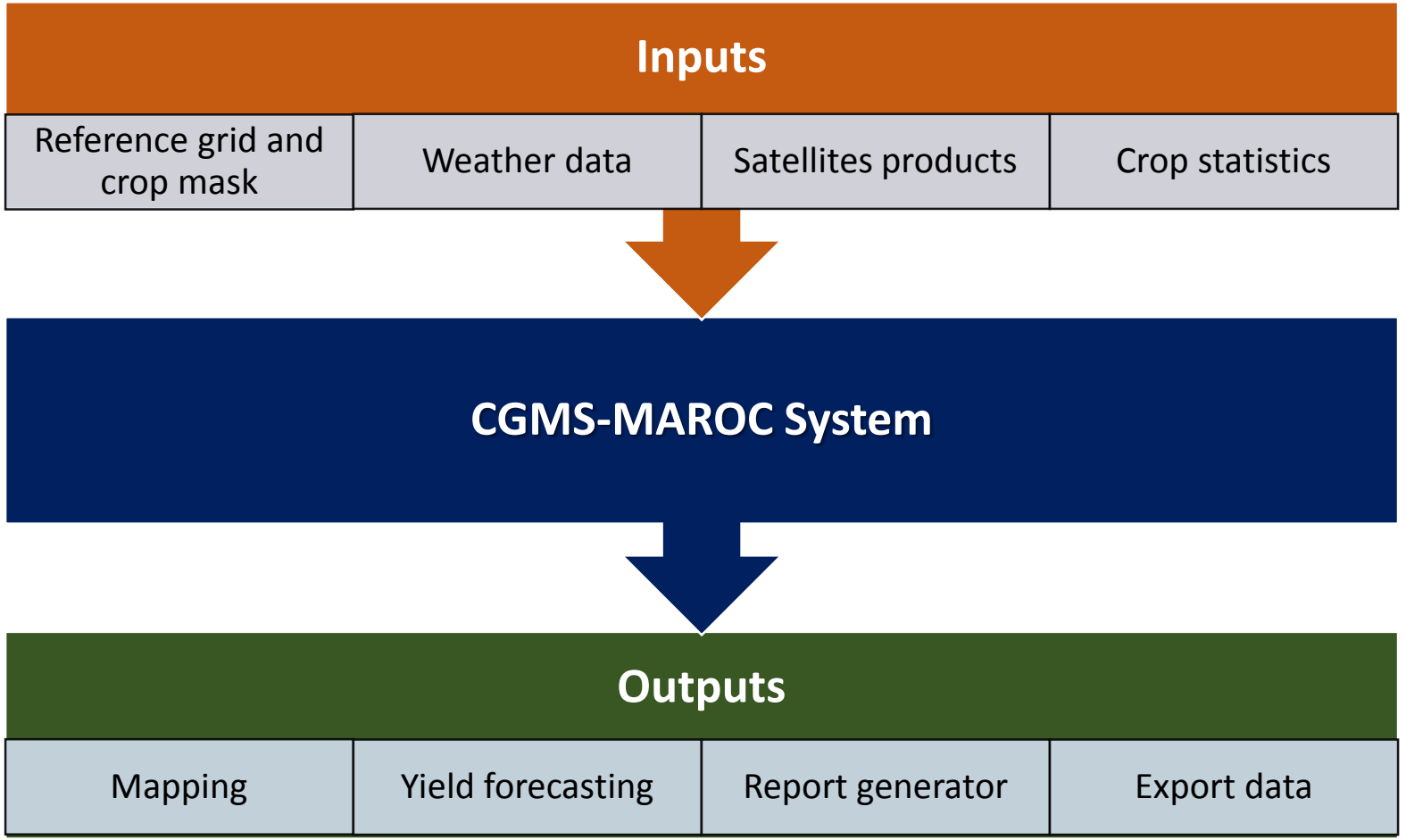
CGMS-Maroc (www.cgms-maroc.ma)

National System for Crop Monitoring

Objectives of the system

- Monitoring the agricultural season
- Support for political decision-making: Anticipating quantities to import
- Index insurance : anticipating farmers repayments
- Area of interest : the hole country

Presentation of the system



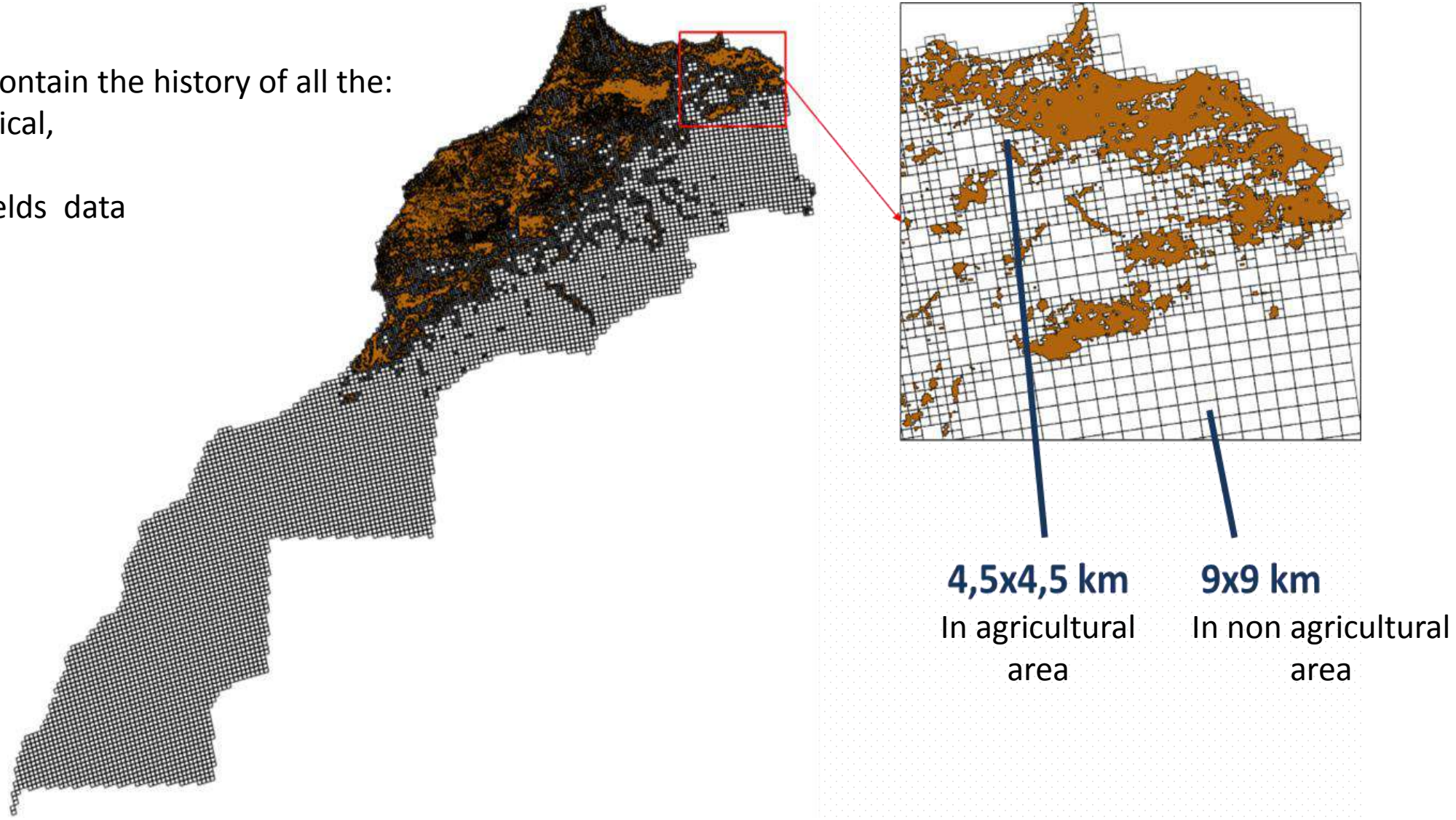
CGMS-Maroc

Data storage

Data grid and agricultural mask

Each grid will contain the history of all the:

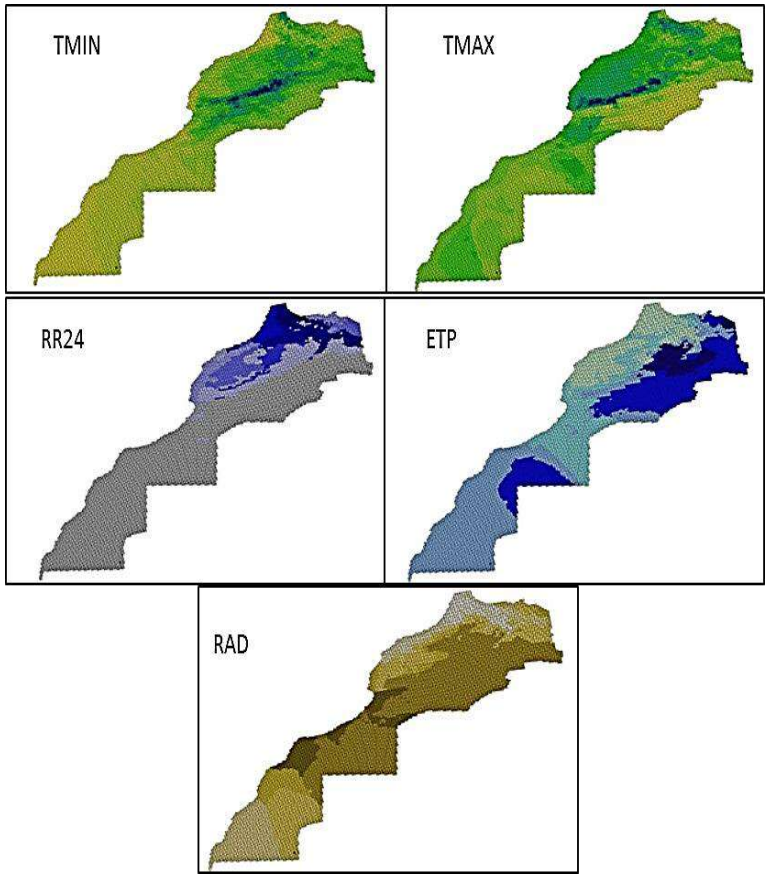
- meteorological,
- satellite
- and crop yields data



Meteorological Data: **Daily** Interpolated Data.

50 Synoptics stations
150 Vigiobs stations

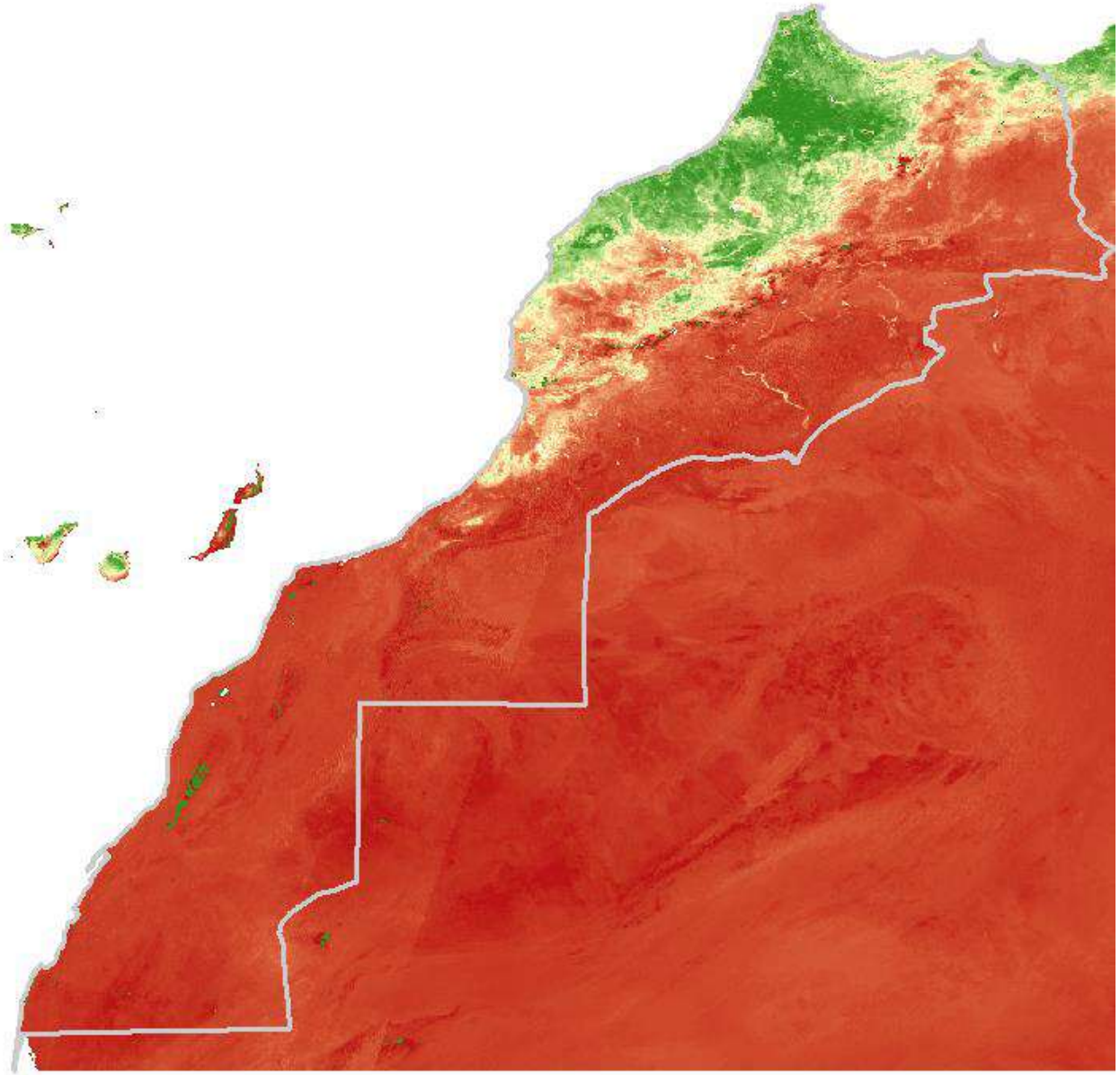
Weather station network



Ex: Grid for 09/02/2014

Satellite **decadal** data

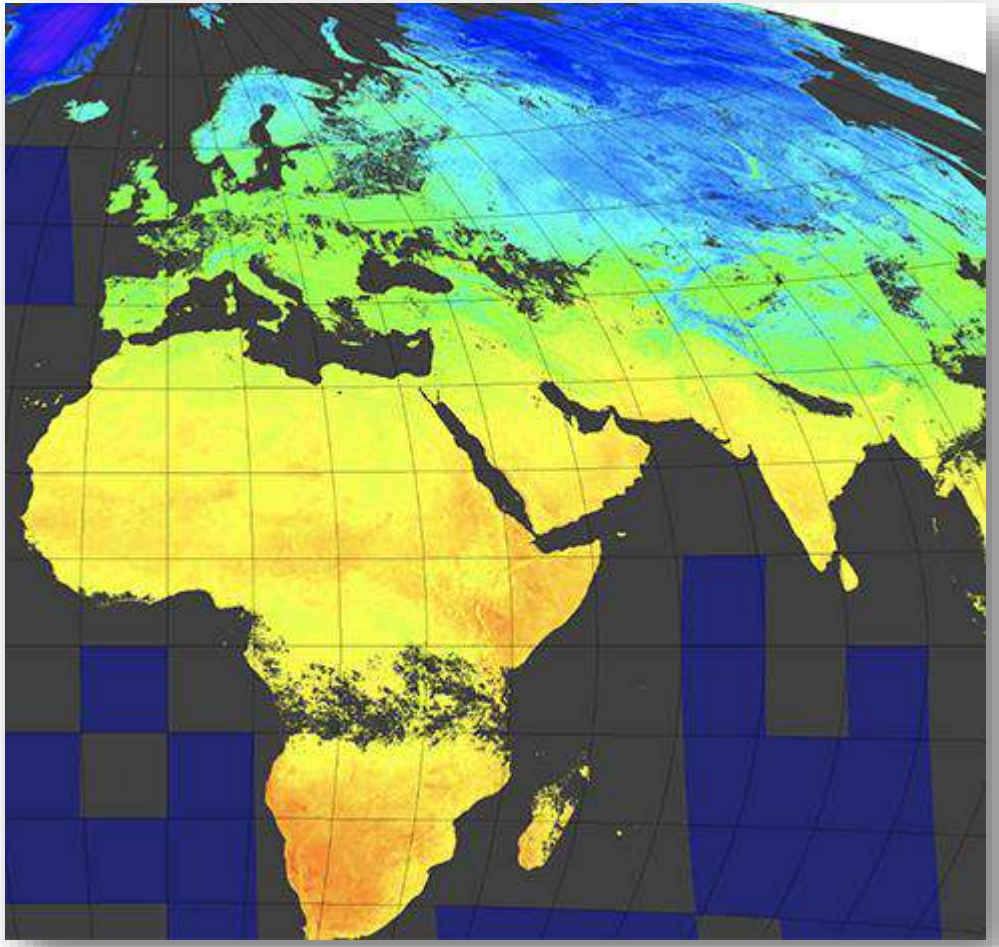
- Vegetation indices derived from satellite images available for free in **Copernicus** Global Land Service (1 km Grid) :
  - NDVI
  - FAPAR
  - LAI



Satellite data

- Estimated agro-climatic data derived from **MODIS** available for free in USGS Land Processes Distributed Active Archive Center (1 km Grid) :
  - **LST** (MOD11A2: Land Surface Temperature)
  - **PET** (MOD16A2: Potential Evapotranspiration)
  - **RET** (MOD16A2: Real Evapotranspiration)
- **RFE :** Satellite-based rainfall: Climate Hazards Group InfraRed Precipitation with Stations (CHIRPS)

All data are automatically download and processed using Python Script

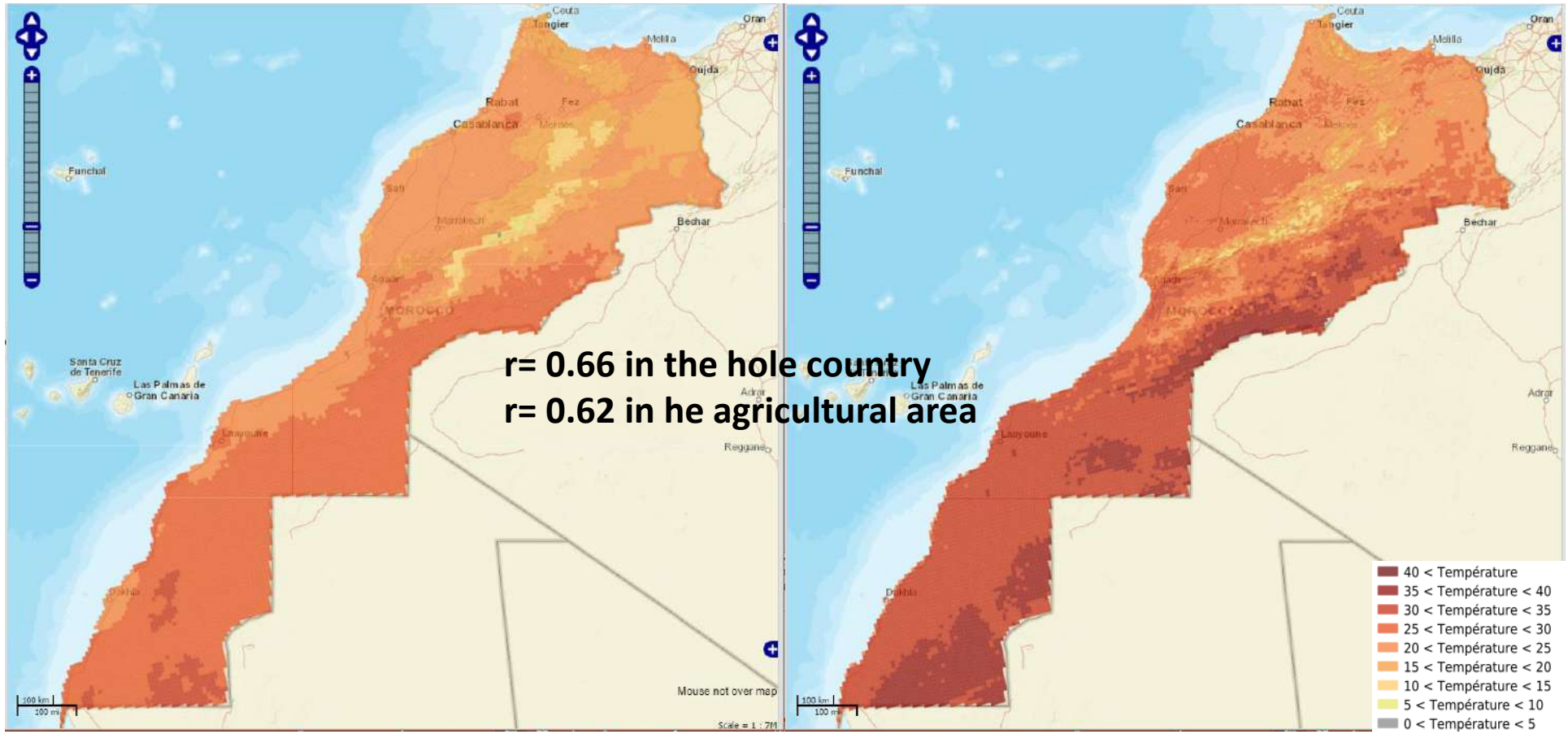


Temperature comparison

The following figures show a comparison between the average maximum temperature taken between the beginning of September 2017 and late April.

Stations terrestres

Satellite

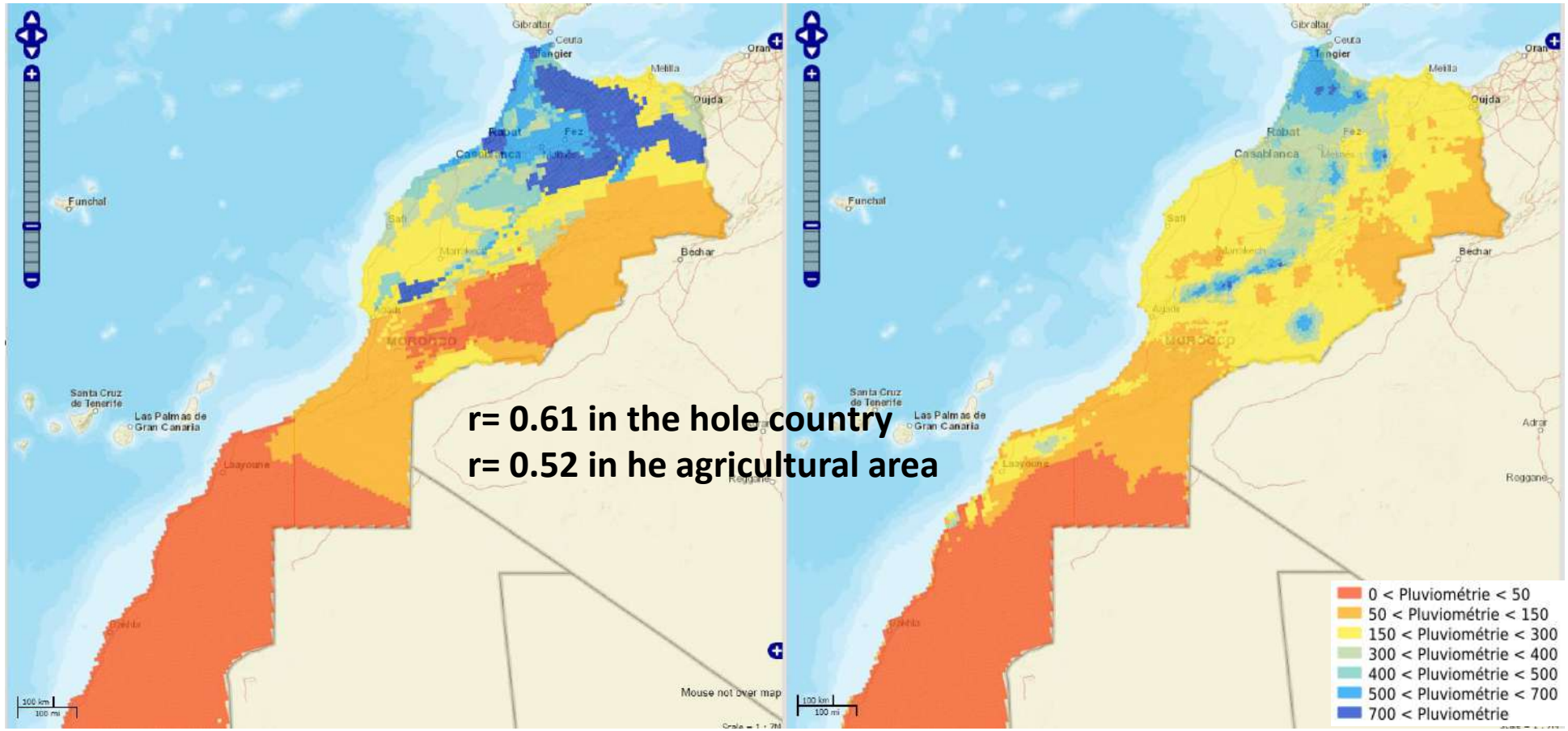


Rainfall comparison

The following figures show a comparison between the total rainfall taken between the beginning of September 2017 and late April.

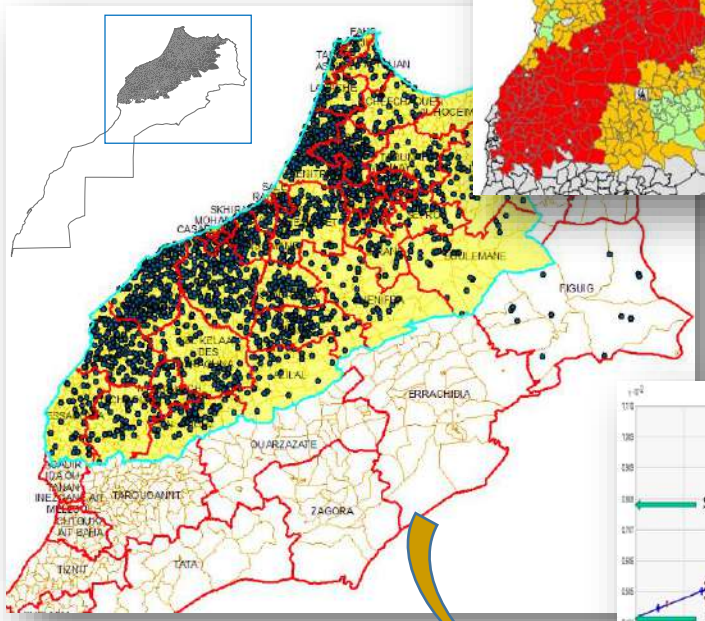
Stations terrestres

Satellite

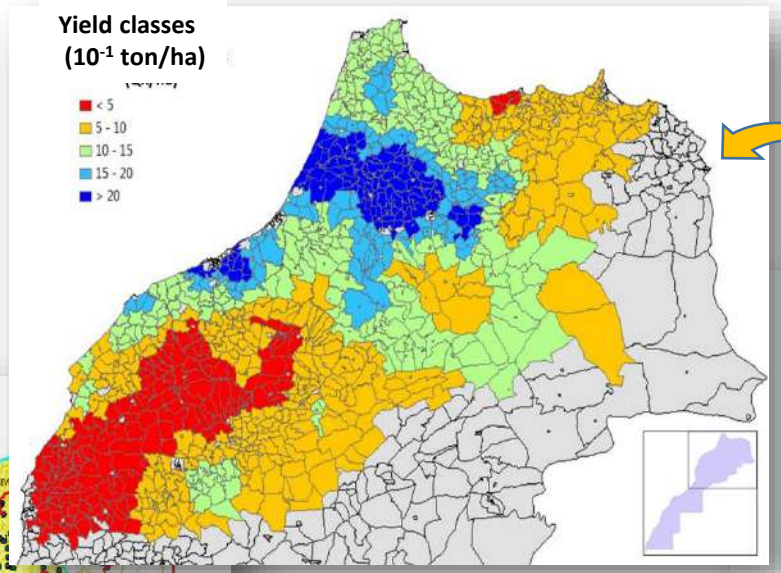


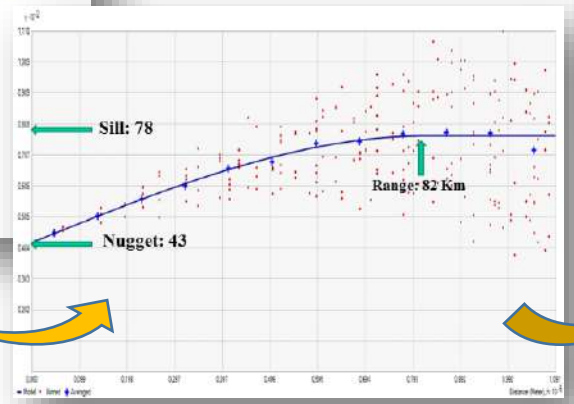
Spatial yield interpolation from sample frames

Area Frame Sampling



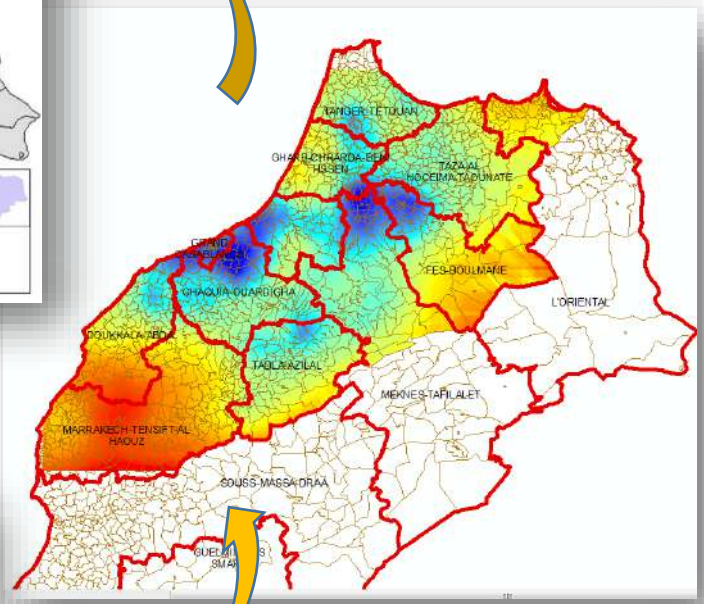
+- 900 districts





Geostatistical model

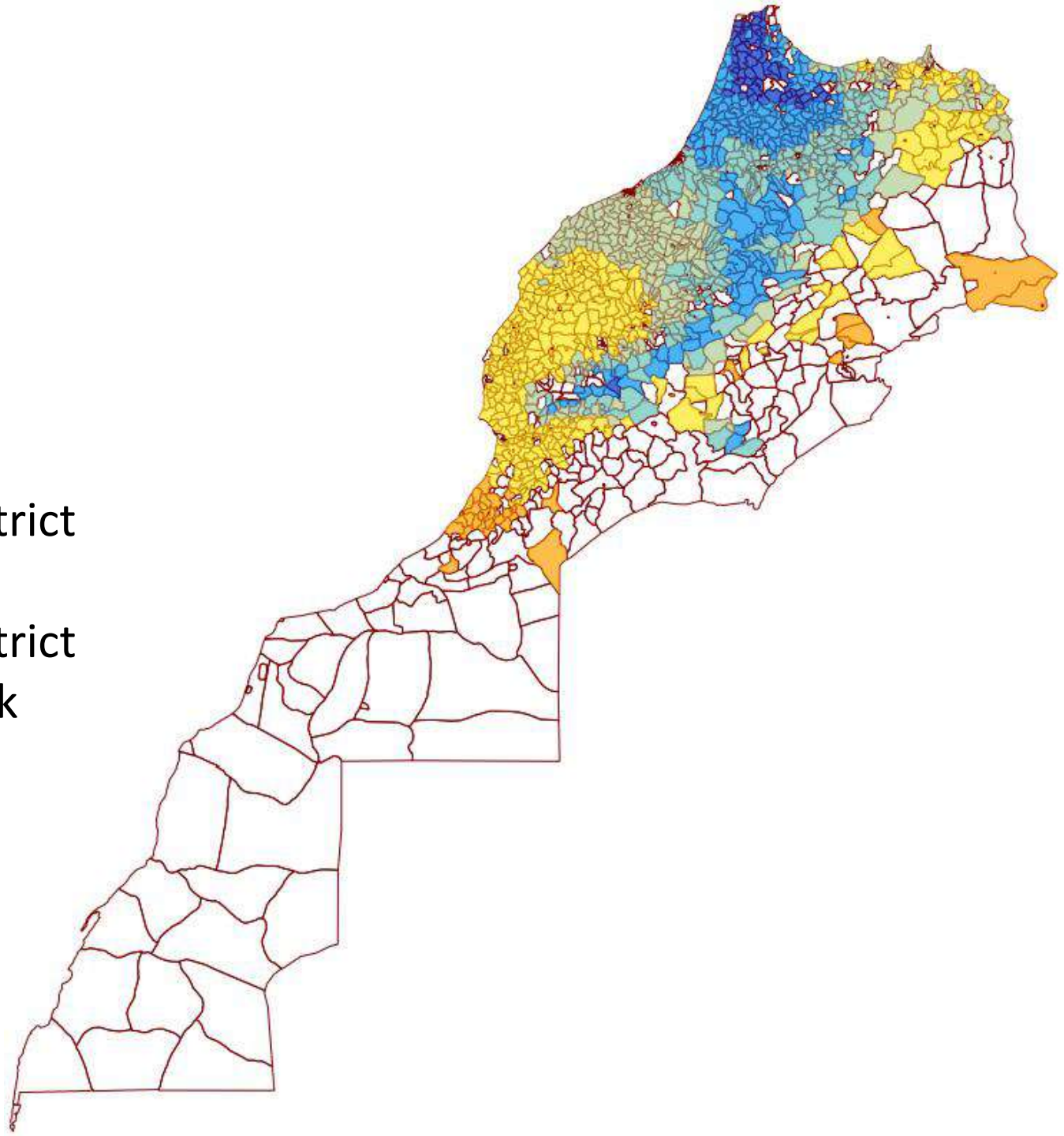
Aggregation at district level



Yield maps (500m GRID) for each cereal and seasons

Data storage

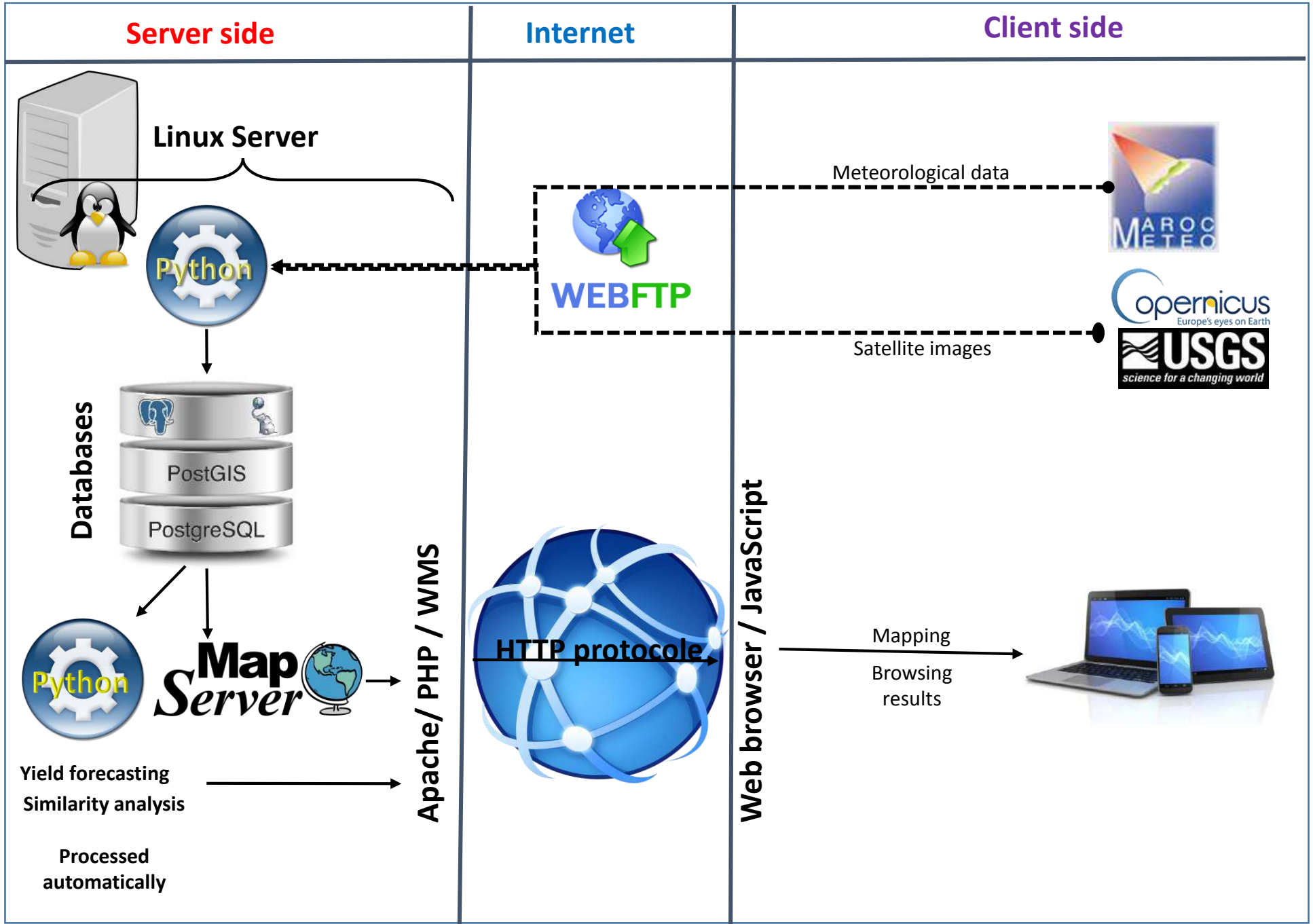
- Data grid
- Data by grid
- District division
- Aggregated data by district
- Agricultural mask
- Aggregated data by district within agricultural mask



CGMS-Maroc

System operation

System architecture (Only open Source tools were used)



CGMS-Maroc

Interface

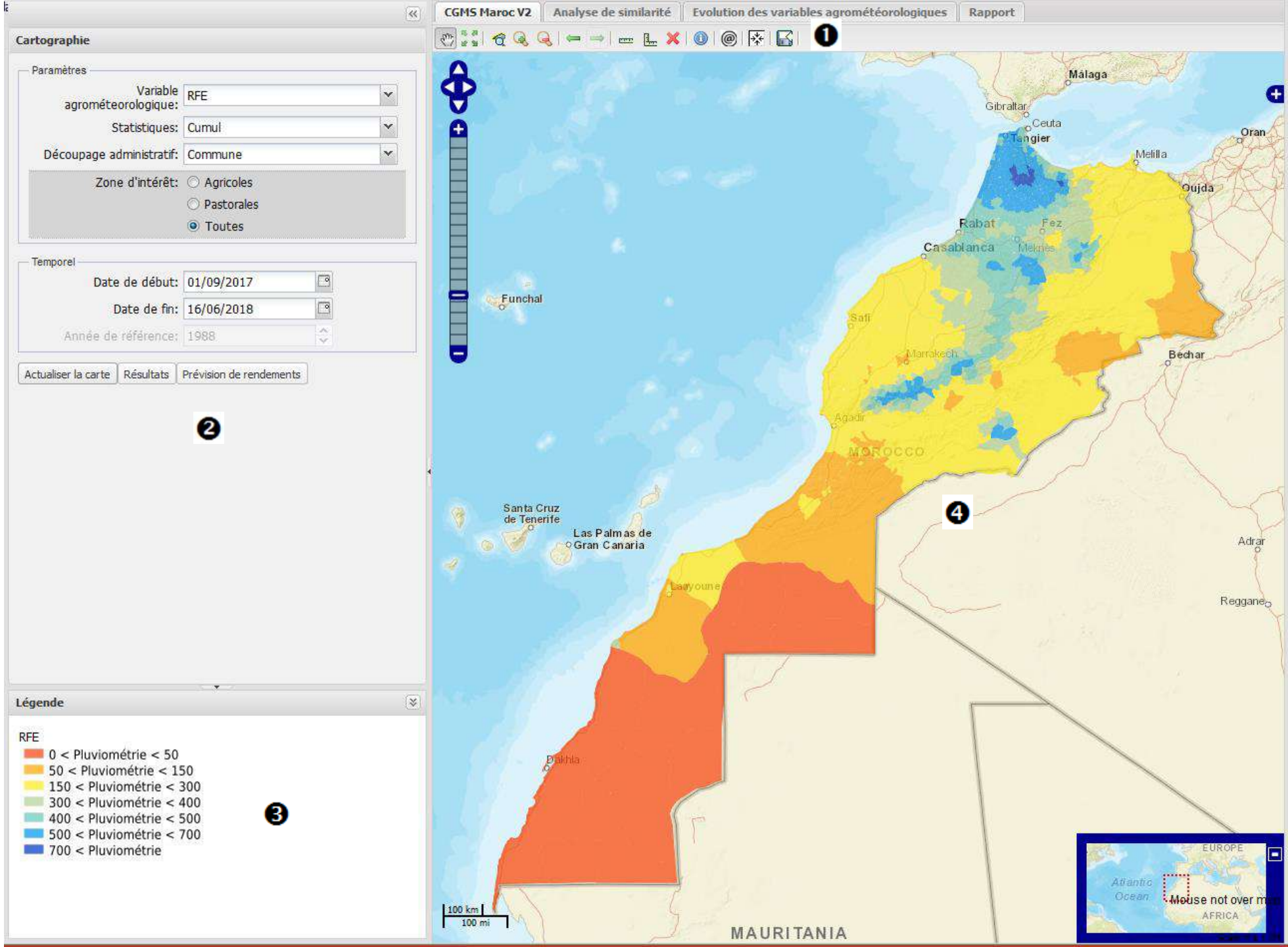
Main interface CGMS-Maroc

Friendly interface with:
(1) toolbar navigation,
(2) query selector,
(3) legend frame,
(4) the map frame

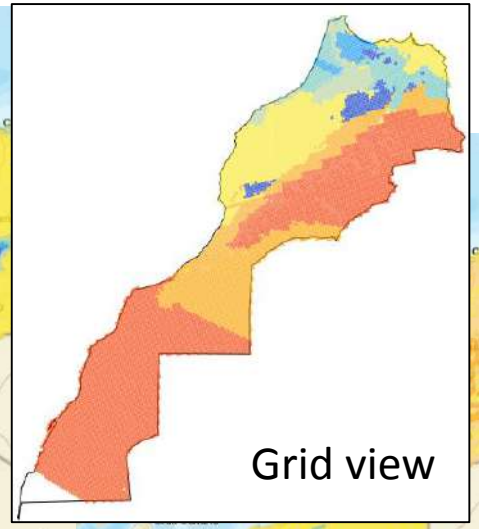
&

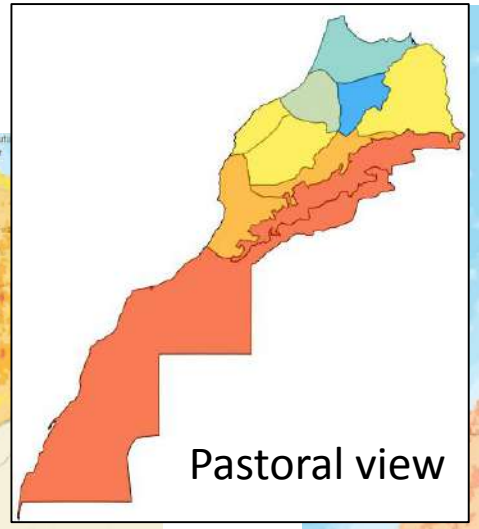
4 Applications

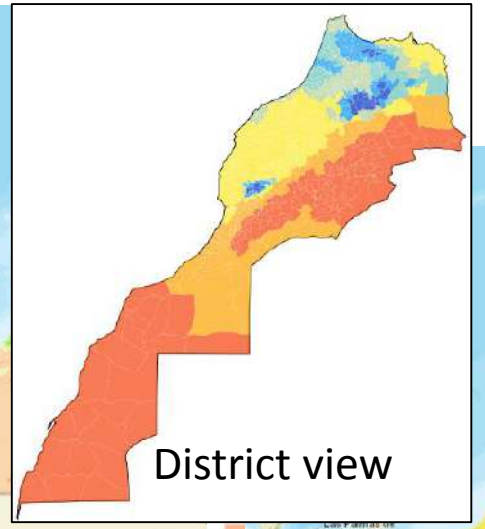
- Mapping
- Daly yield forecasting
- Similarity analysis
- Agro-climatic indicators evolution (data/graph)
- Report generator

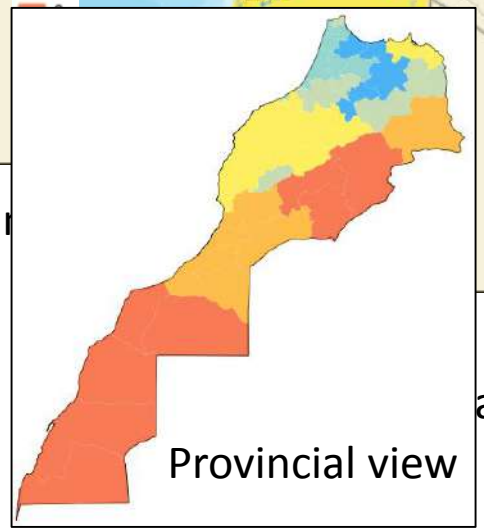


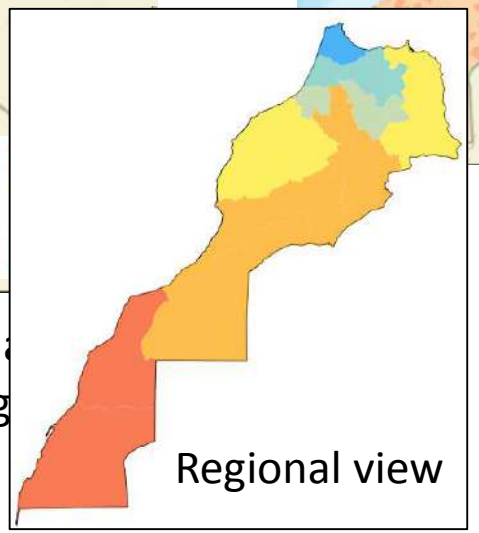
Mapping

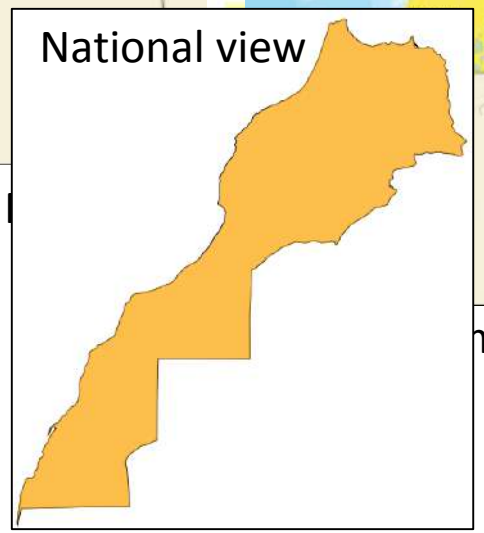












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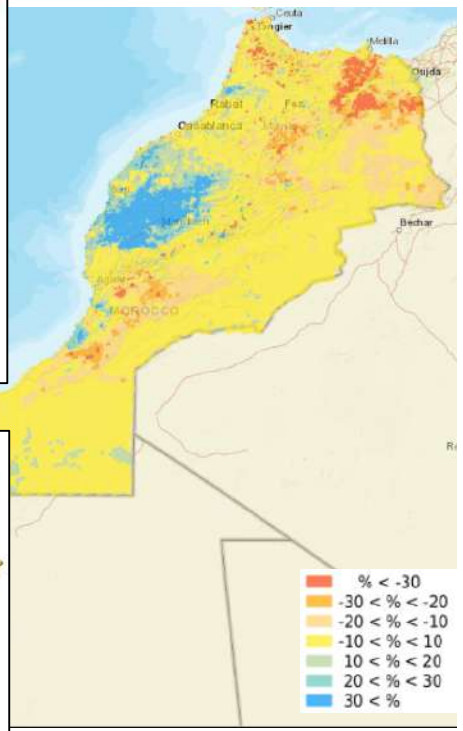
averag



naly with last 10 years average

CGMS-MAROC: Yield forecasting

Performed automatically **each day** from the beginning of February at provincial level with Python script

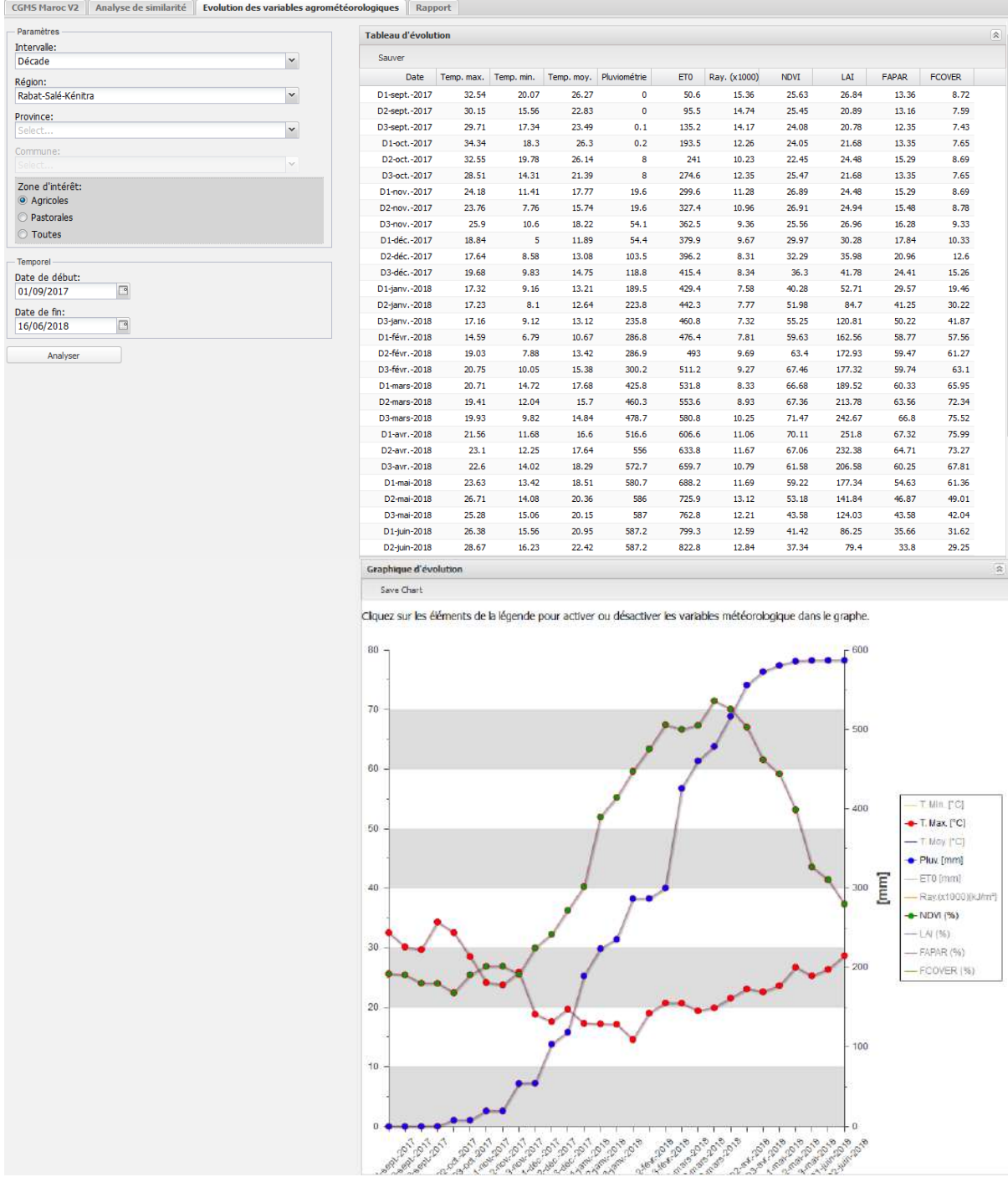
Province	Modèle	N	Prévision rendement LI	Prévision rendement Qx/Ha	Prévision rendement LS	S... Mo... x ...	... ... x 1...	Similarité pluviométrie		Similarité Taux de satisfaction		Similarité NDVI	
								An	Prévision (Qx/Ha)	An	Prévision (Qx/Ha)	An	Prévision (Qx/Ha)
04.-Province:Salé	rdt = -14.7304 + 27.1718*NDVI_Fev_2-42.3041*NDVI_Fev_3 + 65.7886*NDVI_Mar_2,...	16	14.58	19.28	23.98			2010	16	2010	16	2010	16
04.-Province:Sidi Kacem	rdt = -28.5621 + 0.0375*Taux_Satisfaction_P2-0.0164*Regular_S_Pluv + 70.0084*ND...	16	15.66	23.54	31.41			1998	24.23	1998	24.23	1998	24.23
04.-Province:Sidi Slimane	rdt = -25.2765 + 0.032*Taux_Satisfaction_P2 + 59.0402*NDVI_Mar, R²adj=70%	16	11.09	20.48	29.87			1998	25.12	1998	25.12	2003	16.6
04.-Province:Skhirate- Témara	rdt = -10.9404 + 42.5935*NDVI_Mar_2, R²adj=73%	16	12.61	19.28	25.95			2010	15.55	2010	15.55	2013	25.42
05.-Province:Azilal	rdt = -21.7741 + 0.054*Pluv_P2 + 0.1486*Taux_Satisfaction_P1 + 59.6013*NDVI_Moy...	16	0	10.36	20.73			2011	6.71	2011	6.71	2015	12.14
05.-Province:Béni Mellal	rdt = -23.731-0.078*Pluv_P3 + 0.5929*Taux_Satisfaction + 42.8201*NDVI_Avr_3, R²a...	16	2.93	17.86	32.78			2010	12.51	2011	16.69	2013	16.41
05.-Province:Fquih Ben Salah	rdt = -7.0272 + 0.17*Taux_Satisfaction_P1-0.0009*Rayonnement + 37.6271*NDVI_Av...	16	0	12.09	35.8			2011	3.61	2002	6.1	2003	17.48
05.-Province:Khouribga	rdt = 53.219-3.4991*Temp + 42.7219*NDVI_Mar_2-0.1001*Amplitude_NDVI, R²adj=8...	16	12.45	19.91	27.37			2009	20.79	1998	18.54	2003	14.69
05.-Province:Khénifra	rdt = 29.8304-2.6147*Temp-69.9083*NDVI_Fev_2 + 98.1863*NDVI_Mar, R²adj=82%	16	14.3	19.83	25.35			2010	10.67	2011	17.23	2003	12.76
06.-Province:Benslimane	rdt = -20.4685 + 59.6707*NDVI_Mar_2, R²adj=75%	16	11.03	23.96	36.89			2002	11.52	2017	30.56	2010	17.04
06.-Province:Berrechid	rdt = 12.3679-0.0022*Rayonnement + 94.2882*NDVI_Mar_2-56.0388*NDVI_Avr_3, R²...	16	1.91	22.08	42.25			2002	10.86	1998	18.13	2010	16.44
06.-Province:El Jadida	rdt = -55.9091-0.1802*Pluv_P3 + 126.9766*NDVI_Avr_2 + 0.3428*Amplitude_NDVI, R...	16	0	24.44	82.49			2010	13.72	1998	16.69	2010	13.72
06.-Province:Mohammadia	rdt = -35.8422 + 28.7929*NDVI_Fev_2 + 57.4195*NDVI_Mar_2, R²adj=81%	16	15.11	23.1	31.1			2013	25.76	2013	25.76	2010	18.32
06.-Province:Médiouna	rdt = -21.0328 + 540.8807*PNDVI_P3 + 49.1571*NDVI_Fev_2 + 65.3733*NDVI_Mar_...	16	13.92	27.64	41.37			2017	31.19	1998	20.98	2002	12.02
06.-Province:Nouaceur	rdt = -47.5017 + 29.315*NDVI_Fev_2 + 73.7897*NDVI_Mar_2, R²adj=86%	16	15.61	24.06	32.51			2009	30.04	1998	18.58	2002	16.23
06.-Province:Settat	rdt = 32.5382-0.0026*Rayonnement + 80.4171*NDVI_Mar_2-43.5697*NDVI_Avr_1, R²...	16	13.53	23.91	34.29			1998	18.7	1998	18.7	2010	13.33
06.-Province:Sidi Bennour	rdt = 170.5952-0.0018*Rayonnement-9.7392*Temp + 68.7976*NDVI_Mar, R²adj=79%	16	8.45	29.94	51.42			1998	15.78	1998	15.78	2011	21.46
07.-Province:Al Haouz	rdt = -13.9018 + 54.7101*NDVI_Avr_1, R²adj=82%	16	7.39	11.39	15.39			2011	1.72	2011	1.72	2009	16.73
07.-Province:Chichaoua	rdt = -1.9543 + 0.0251*Pluv_P3 + 62.8803*NDVI_Mar_2-54.0399*NDVI_Avr_3, R²adj...	16	6.47	7.81	9.15			2011	4.39	2011	4.39	1998	6.92
07.-Province:El Kelâa des Sraghna	rdt = 21.1396 + 0.0762*Taux_Satisfaction_P1-2.0809*Temp_P3 + 42.1885*NDVI_Mar...	16	7.06	20.32	33.58			1998	11.16	2002	5.22	2010	9.23
07.-Province:Essaouira	rdt = -5.9056 + 0.137*Taux_Satisfaction_P1-19.867*NDVI_Mar_3 + 34.3314*NDVI_Av...	16	1.87	3.65	5.43			2003	7.41	1998	9.13	2011	9.12
07.-Province:Marrakech	rdt = -12.4859 + 35.7884*NDVI_Fev_2 + 36.4922*NDVI_Mar-124.6901*NDVI_STD, R²...	16	9.36	12.03	14.69			2011	4.78	2011	4.78	1998	6.57
07.-Province:Rehamna	rdt = 65.7033-0.0007*Rayonnement-3.5967*Temp + 38.6444*NDVI_Mar_2, R²adj=89%	16	13.46	18.91	24.36			1998	10.32	2011	7.36	1998	10.32
07.-Province:Safi	rdt = 27.3208 + 0.0247*Regular_S_Pluv-2.0514*Temp_P2, R²adj=67%	16	0	6.11	14.27			1998	12.51	1998	12.51	1998	12.51

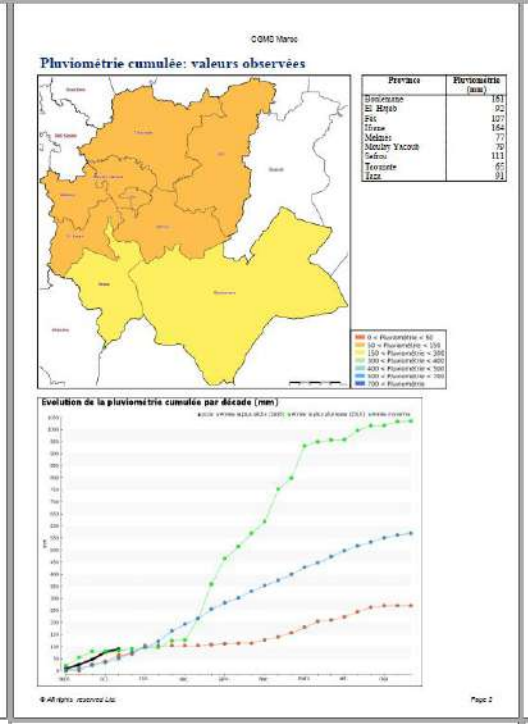
Similarity analysis

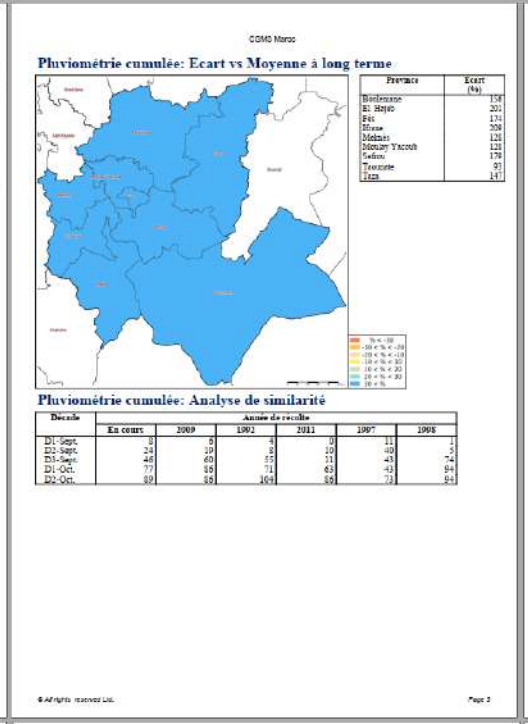
Rapid characteristics of the cropping season, by comparing the similarity of the past seasons to the current one, from an agro-climatic point of view.

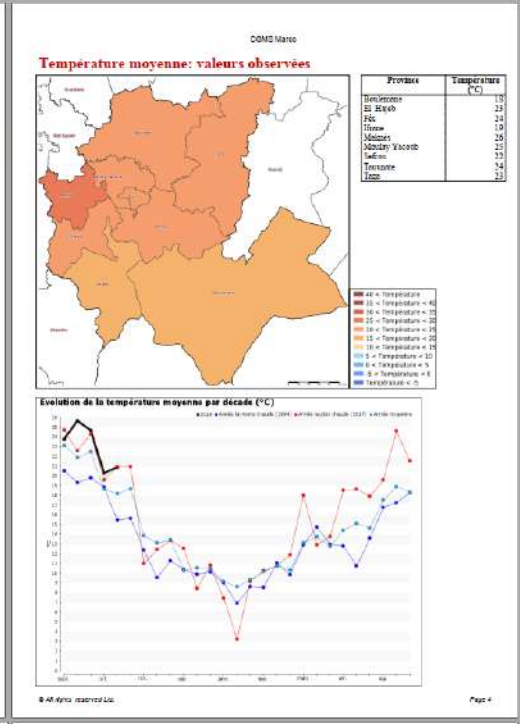
Agro-climatic indicators evolution (data/graph)

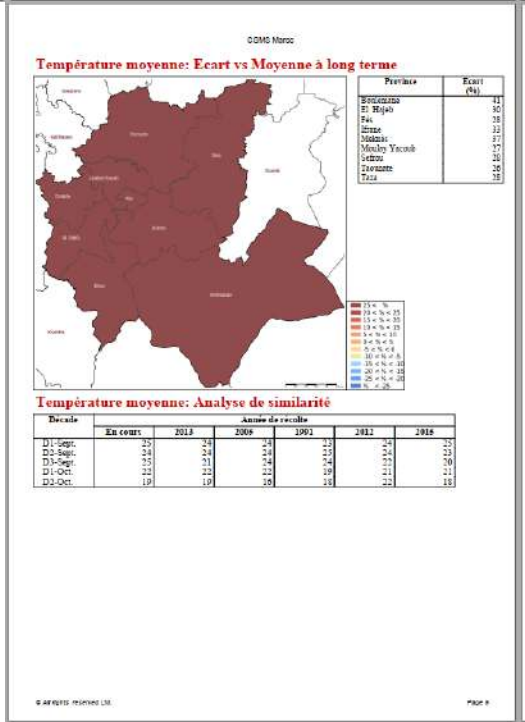
This feature allows displaying in tables and graphs the evolution of agro-meteorological indicators.

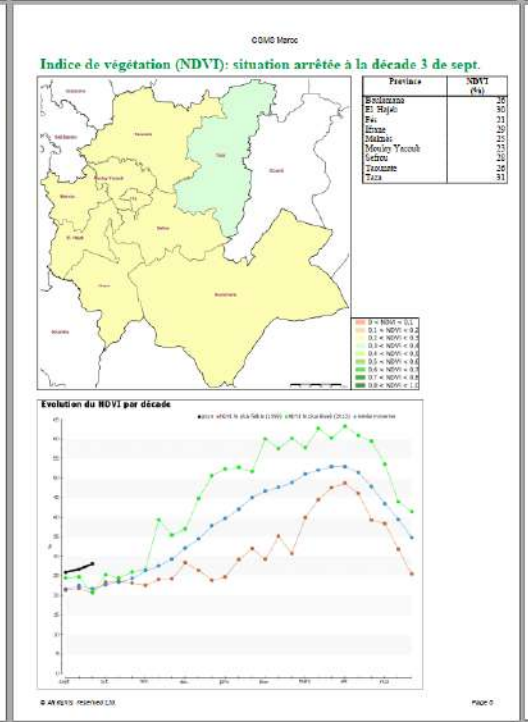


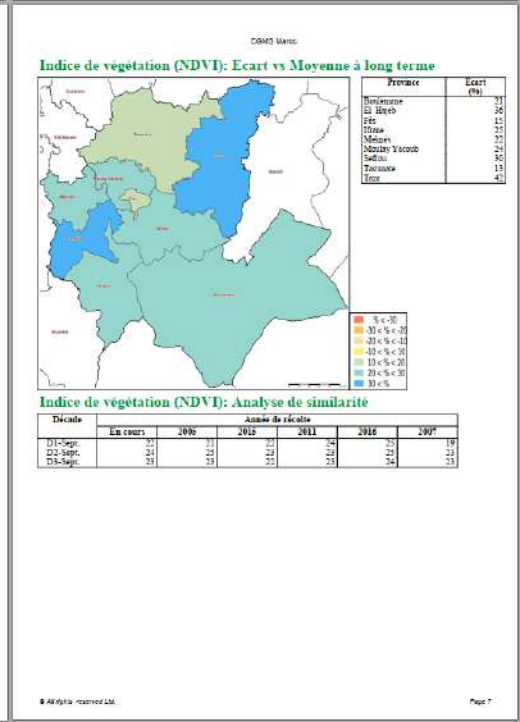












Reporting feature

Automatic generation of reports at different levels:

- National
- Regional
- And Provincial

Including **rainfall**, **temperature** and **NDVI** evolution and anomalies

Operational cereal yield forecasting in Morocco

Purpose of the study

1. Compare the two approaches of Machines Learning:
    - Statistic: Multiple linear regression
    - Learning: Random Forest and Boosted Tree
2. Quantify the contribution of satellite data in crop yield prediction by comparing models based on the use of:
    - Agro-climatic data from Earth observation
    - Vegetation indices from Copernicus Global Land Service
    - Estimated agro-climatic data derived from MODIS

Data characterization

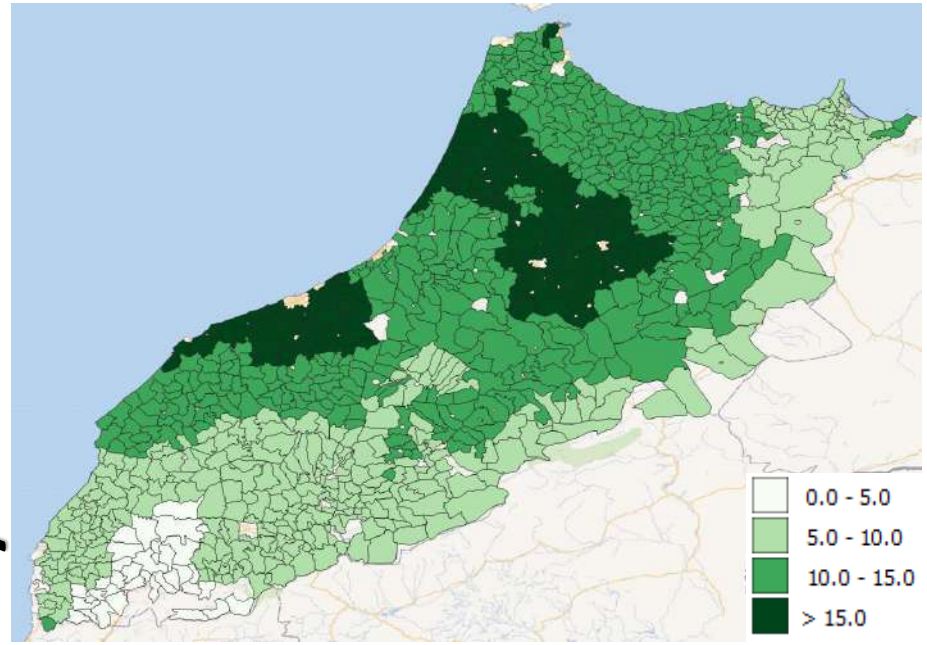
- **1** dependent variable : yield
- **140** predictors
  - **4** geographical information
  - **58** meteorological
  - **42** vegetation indices
  - **69** Estimated agro-climatic
- 35890 lines
  (14 years observation, 3 cereals and ≈ 900 districts)

Simulation

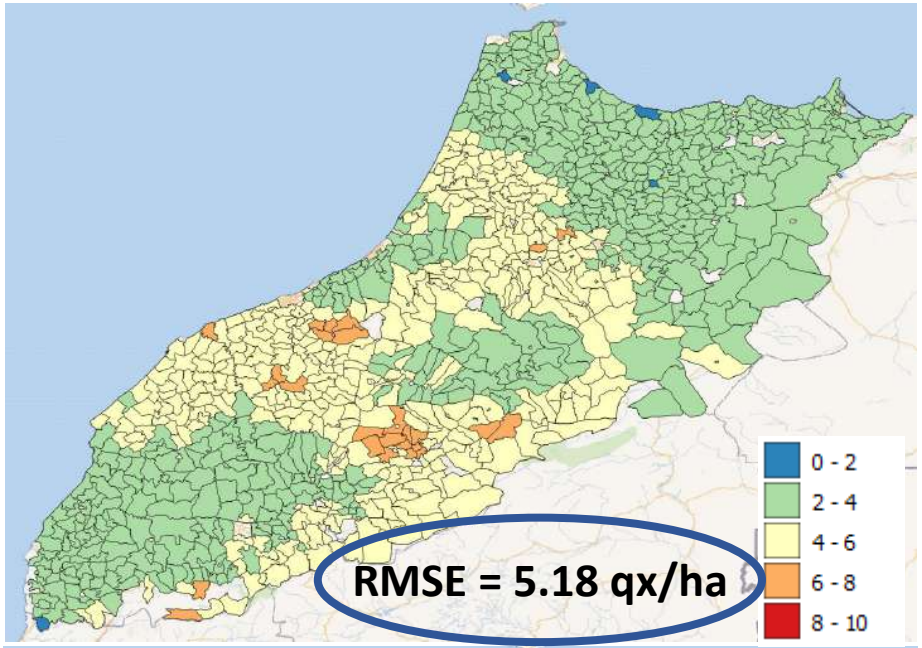
- Automation scripts have been developed for the three forecasting techniques selected for this study:
  - Multiple linear regression
  - Random Forest
  - Boosted Tree
- To perform calculations we
  - For each available growing season,
    - The data was separated into three different subsets to ensure model accuracy:
      1. Testing data that correspond to the growing season been analyzed,
      2. Validation data (fraction of 20%)
      3. Training data (remaining data).
    - The training data was used to build model. Once both the training and validation prediction results are similar to the observed, we use the model to predict the yield for the test data subset.

Absolute error: Comparison between prevision models

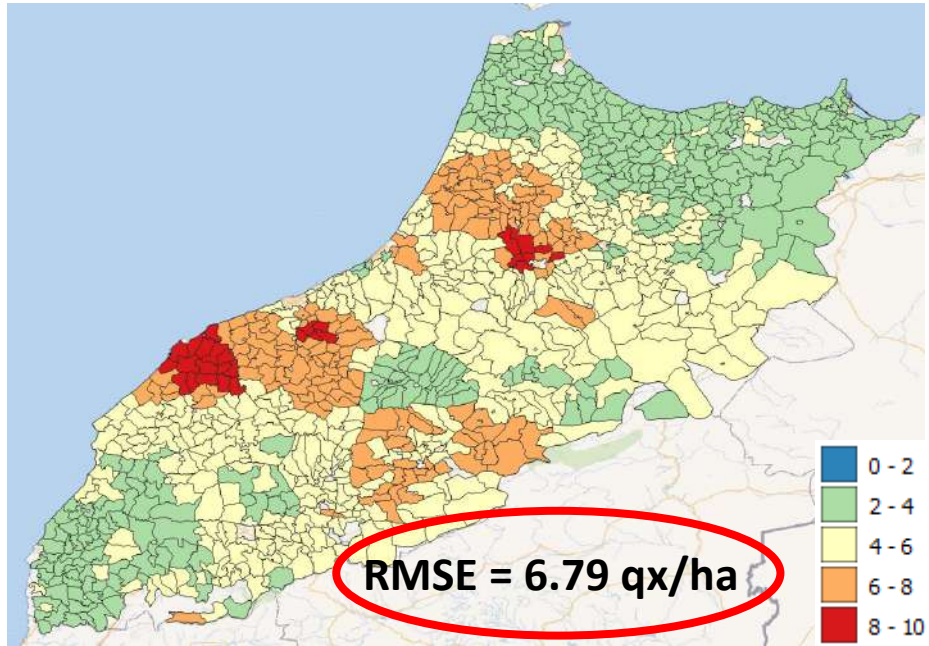
Average observed yield 2002-2017



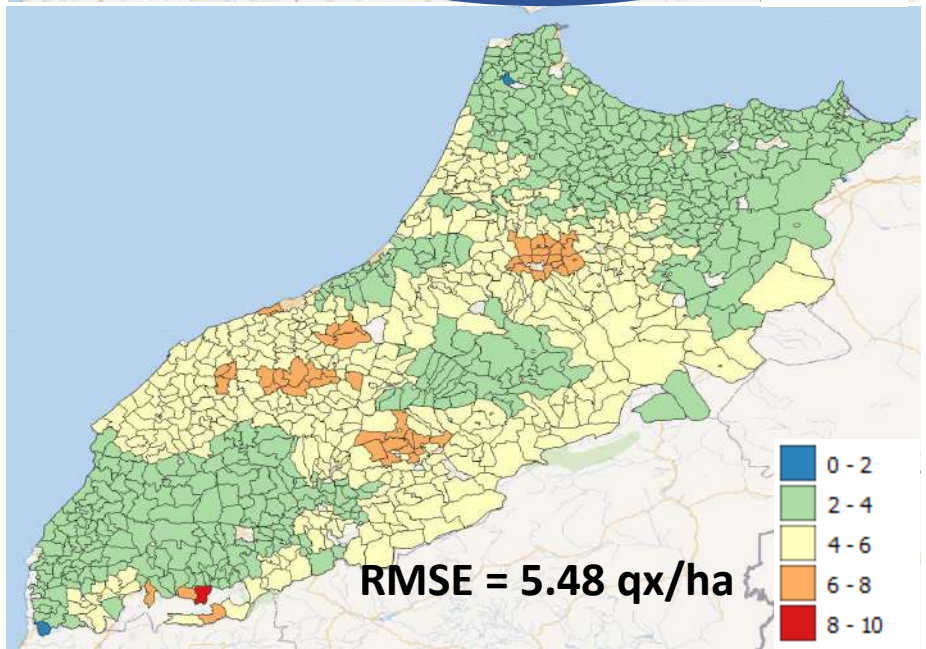
Random Forest



Linear Model

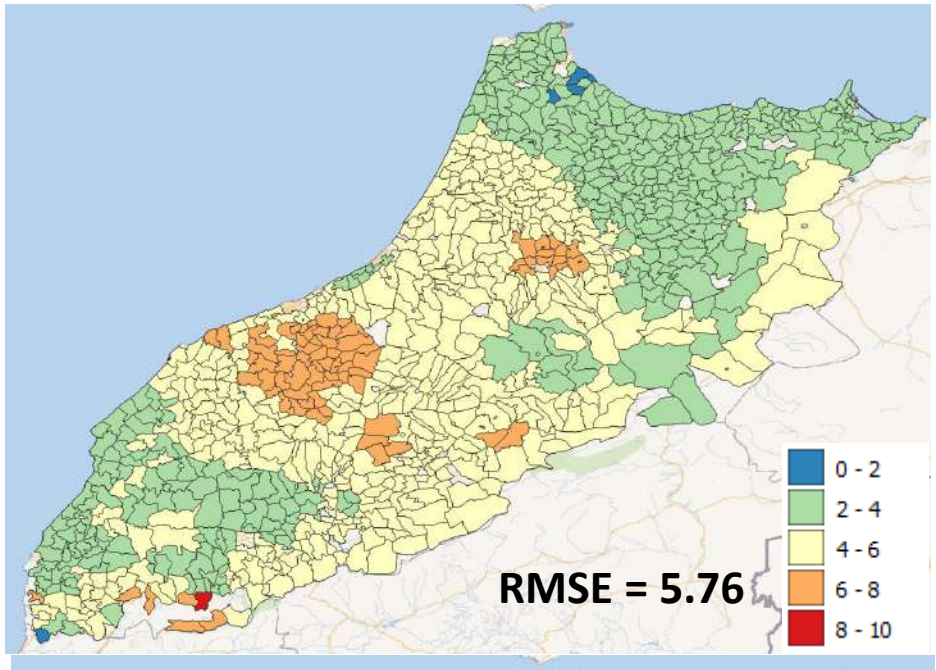


Boosted Tree

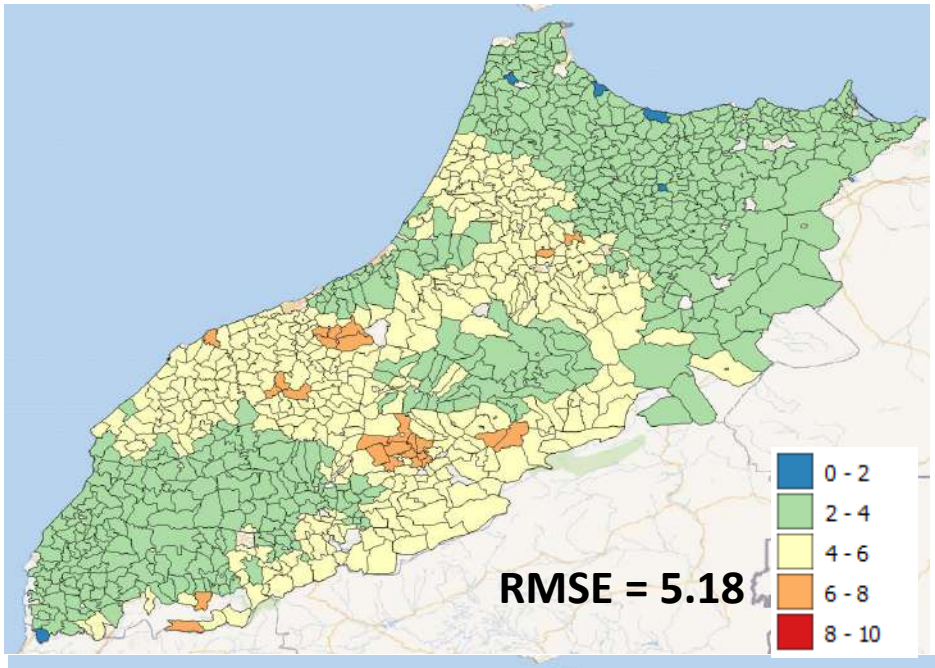


Absolute error: Comparison between data sources

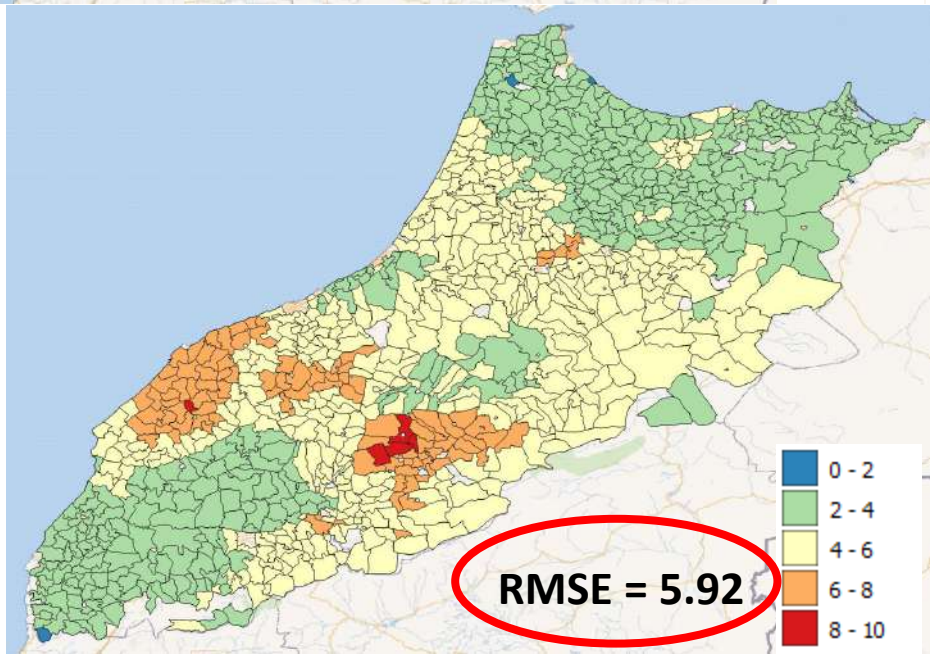
Weather data



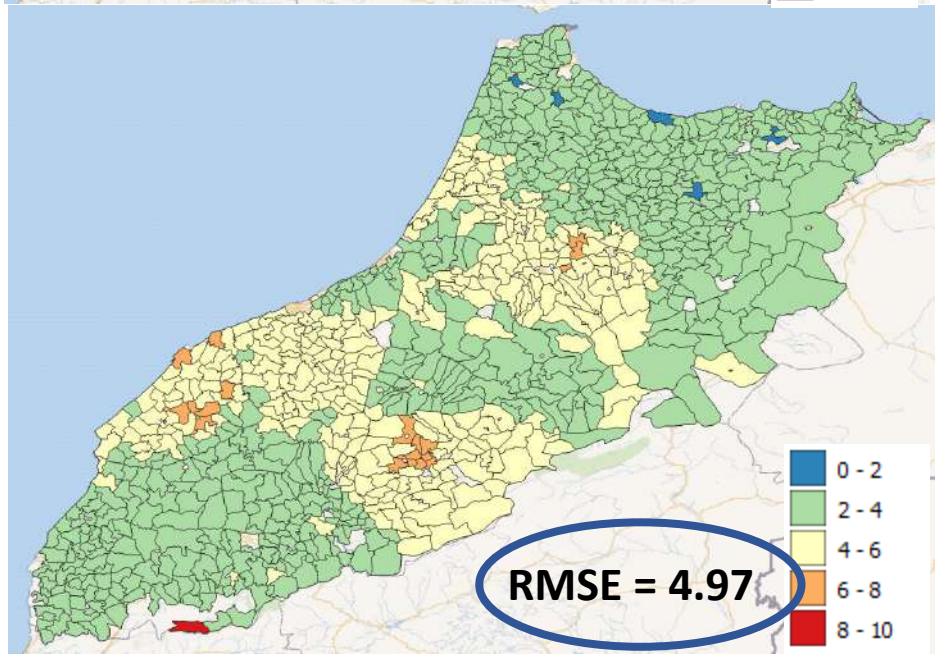
Weather & Satellite



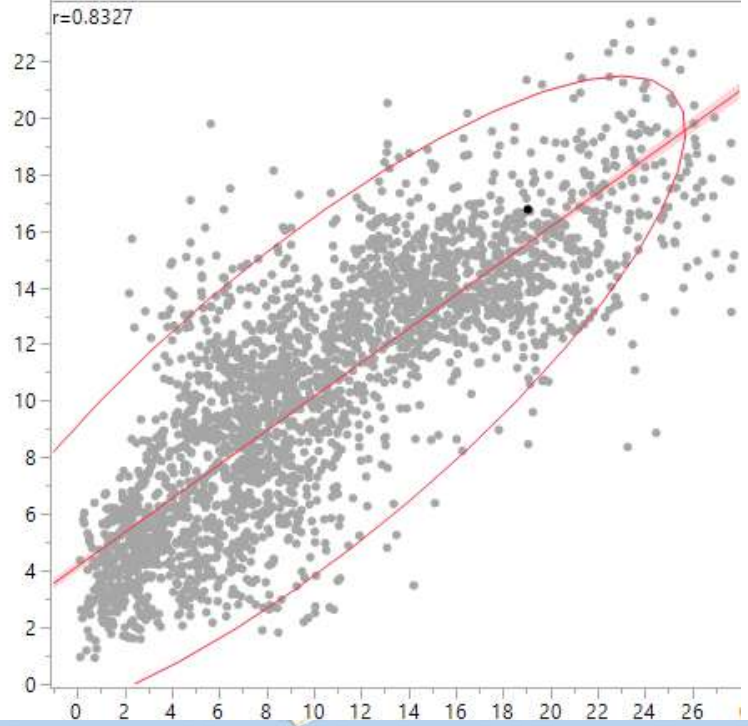
Vegetation indices (Satellite)



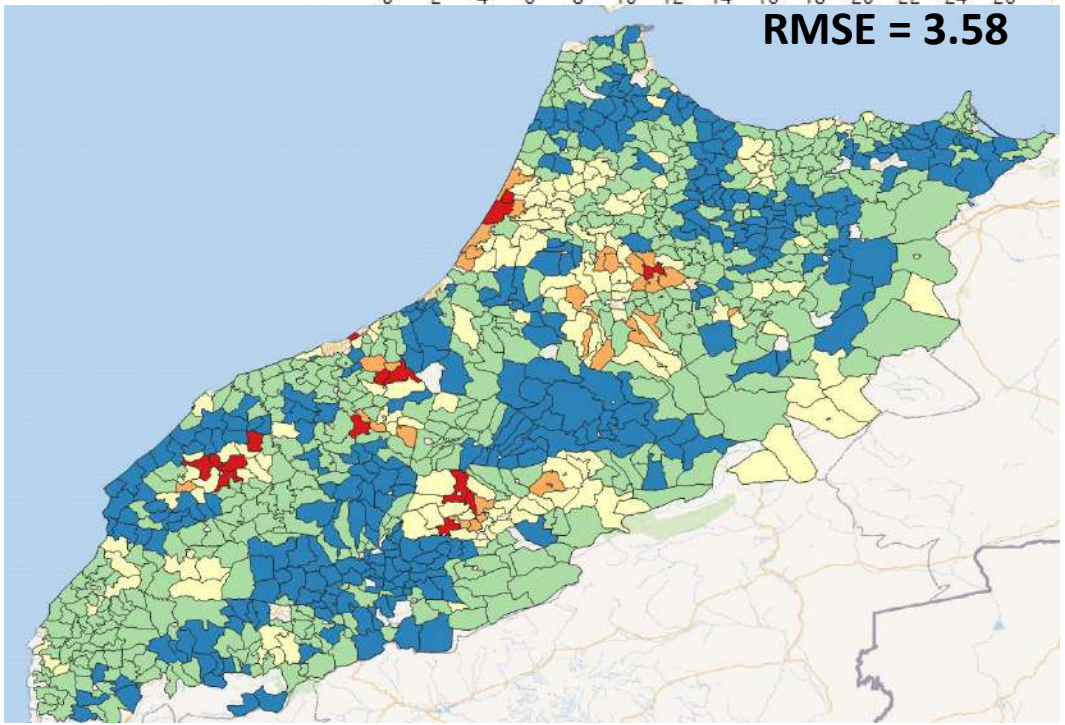
Satellite & Weather estimated from Sat



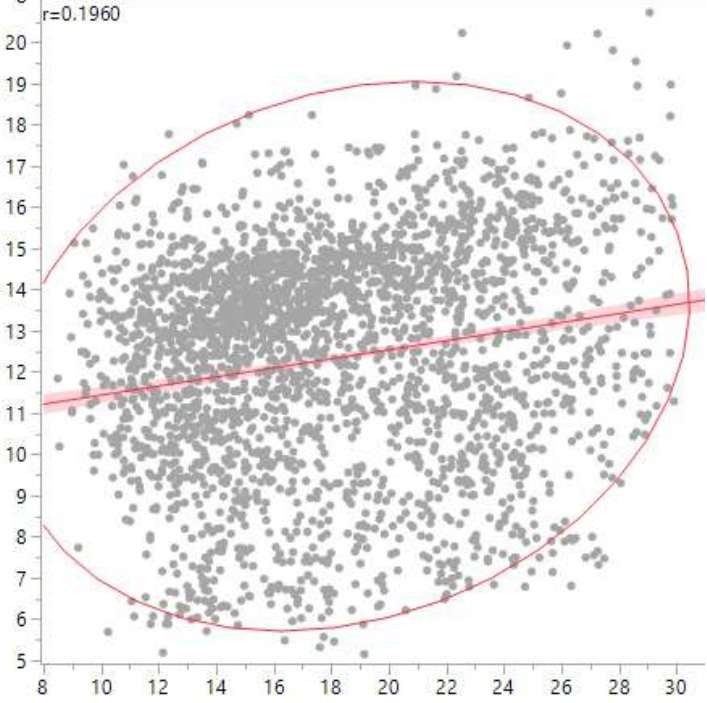
Case study

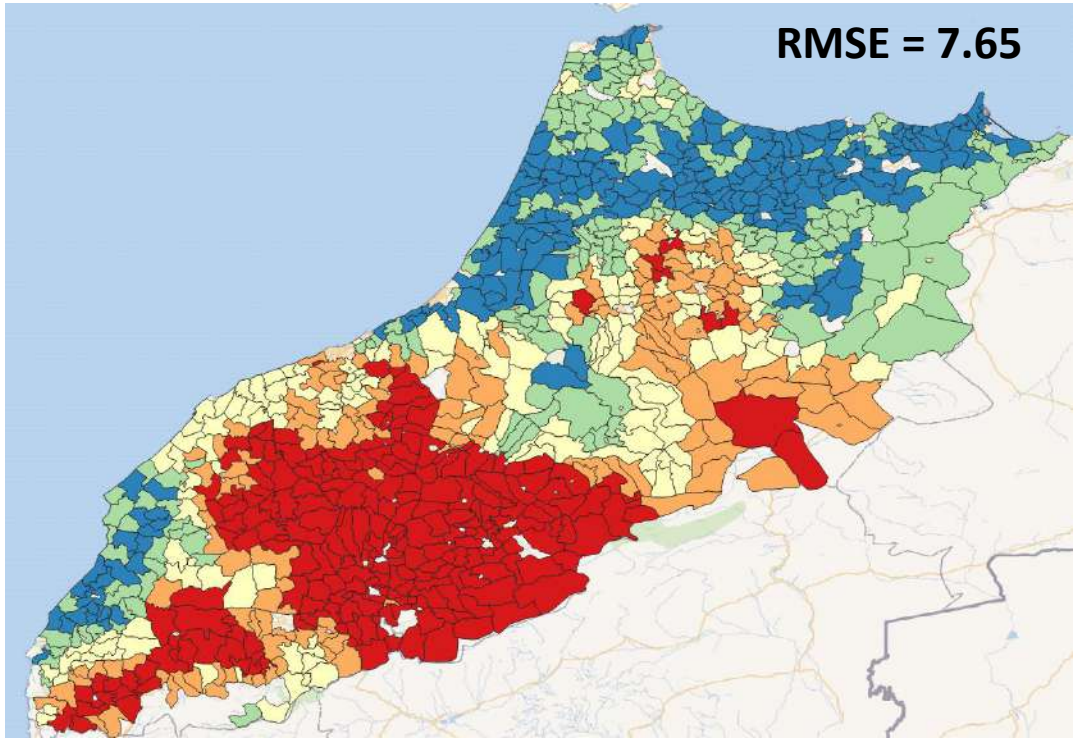


2014 year
Average production



2009 year
High production



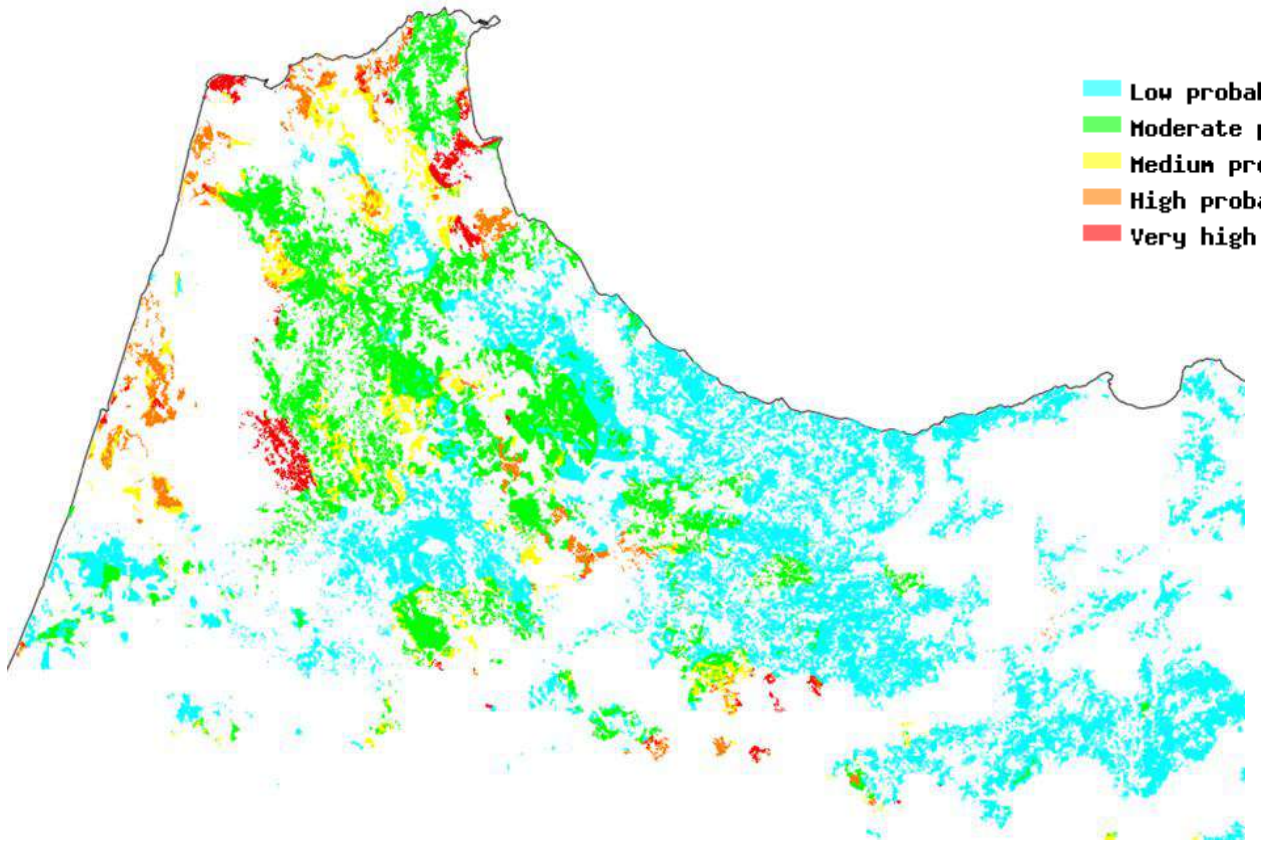


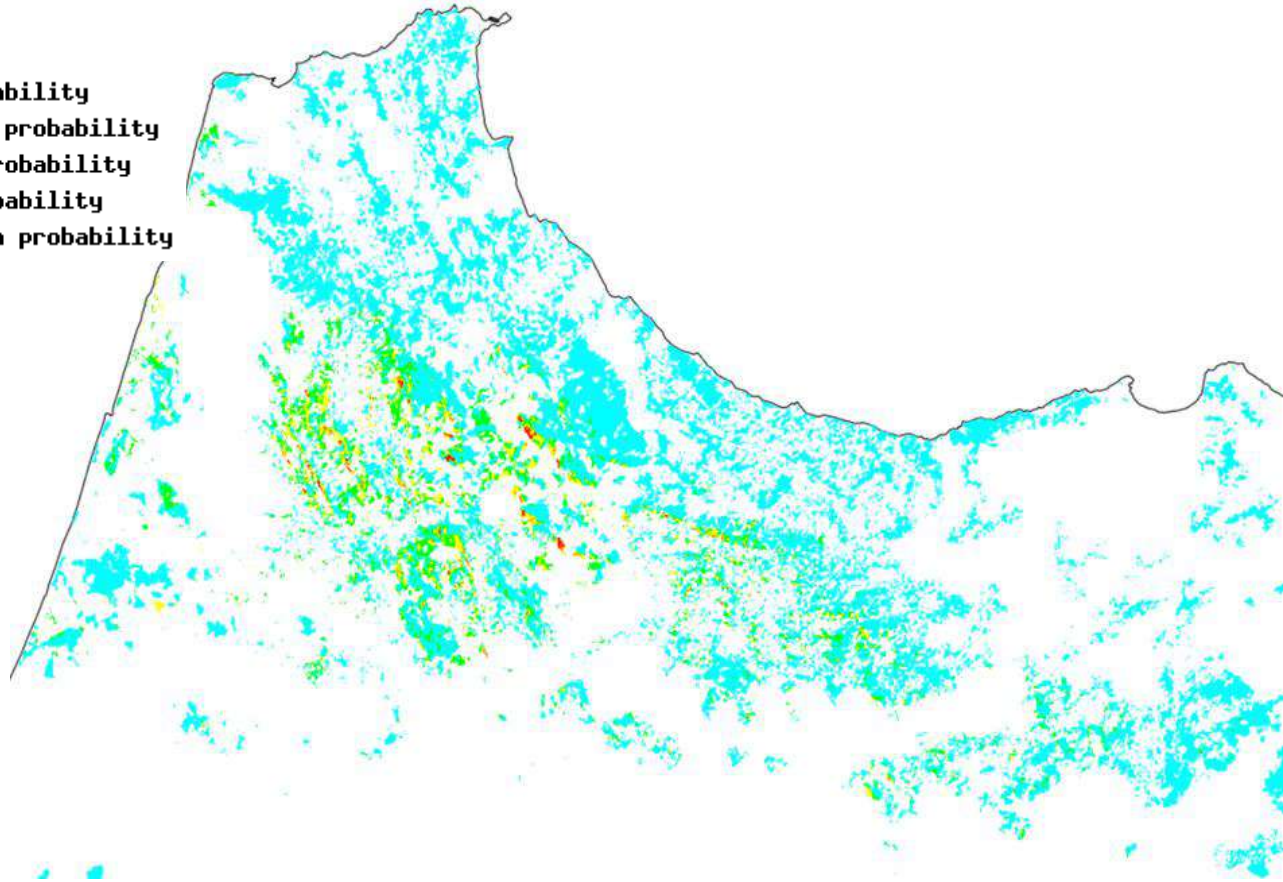
Future Improvement

- Land cover form last census (irrigated / rainfed)
- Land cover from Copernicus
- Move from 1 km to 300 m or 100 m with more accurate land cover
- Use phenology information derived from satellite images
- Use multi-model approach (like use AquaCrop output as predicator)

Producing Forest Fire Risk Maps at (day-1 and day-2)

By combining statics (Land & Forest) and dynamics (Satellite & meteorological) maps.

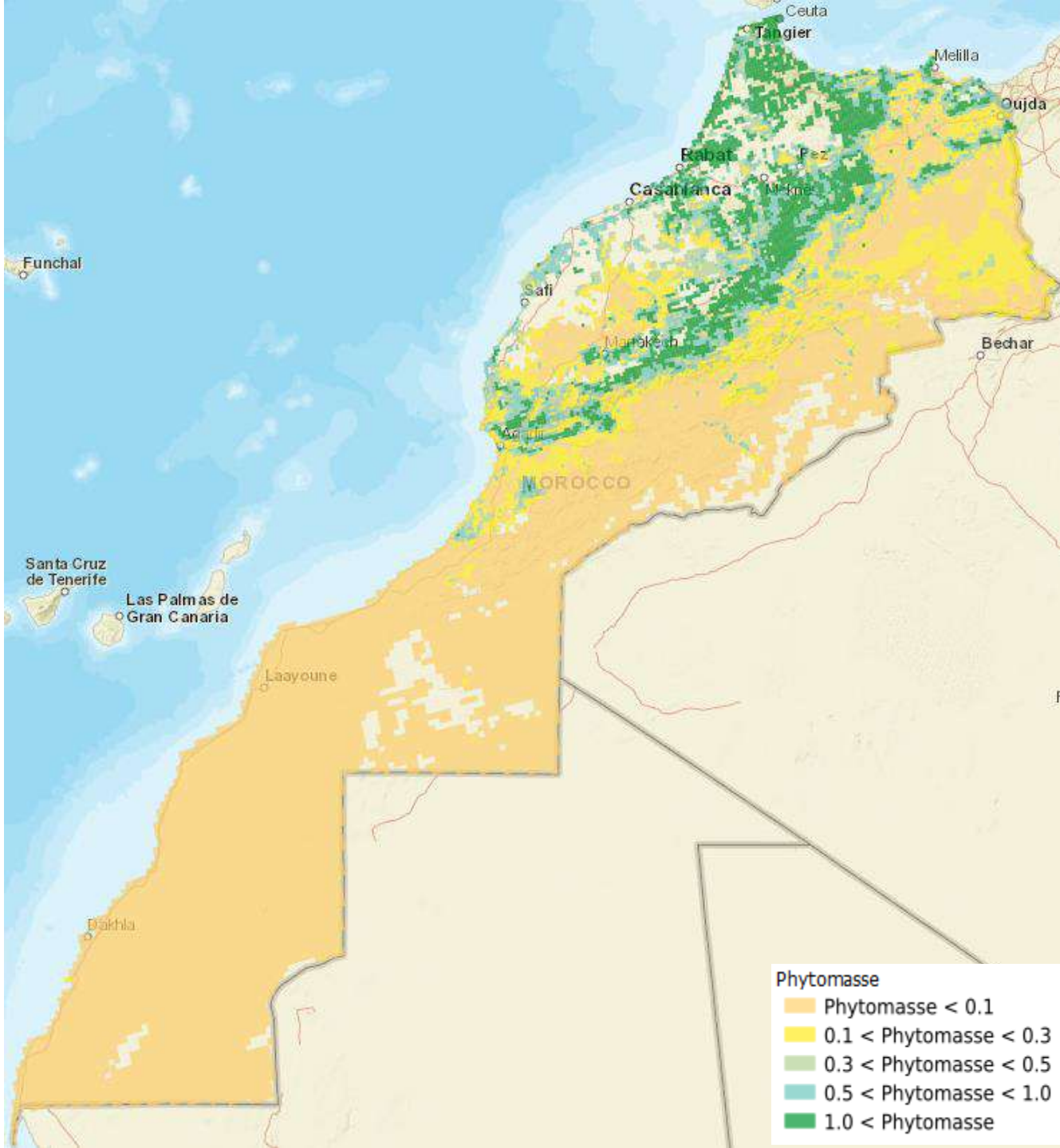




Ignition risk map

Propagation risk map

Estimation of Phytomass production in Rangelands.



Thank you