

WP8 - SCIENTIFIC AND TECHNICAL VALIDATION OF PRODUCT AND SERVICES

Speaker: Monica Pepe (CNR-IREA)

Task 8.1 Leader: Ioannis Gitas (AUTH)

Task 8.2 Leader: Manuel Campos-Taberner (UVEG)

Task 8.3 Leader: Ignacio Miralles (UJI)

Main Contributors *(In no particular order):*

Dimitrios Katsantonis (DEMETER)

Dimitris Stavrakudis (AUTH)

Goncal Grau, Javier Garcia Haro, Beatriz Martínez (UVEG)

Francesco Holecz, Massimo Barbieri, Luca Gatti (SARMAP)

Filomena Romano, Elisabetta Ricciardelli, Mariassunta Viggiano (IMAA)

Alberto Crema, Francesco Nutini, Luigi Ranghetti, Lorenzo Busetto, Daniela Stroppiana, Mirco Boschetti (IREA)

Roberto Confalonieri, Simone Bregaglio, Valentina Pagani, Tommaso Stella (UMIL)

Nacho Miralles, Carlos Granell, Sven Casteleyn, Sergi Trilles (UJI)

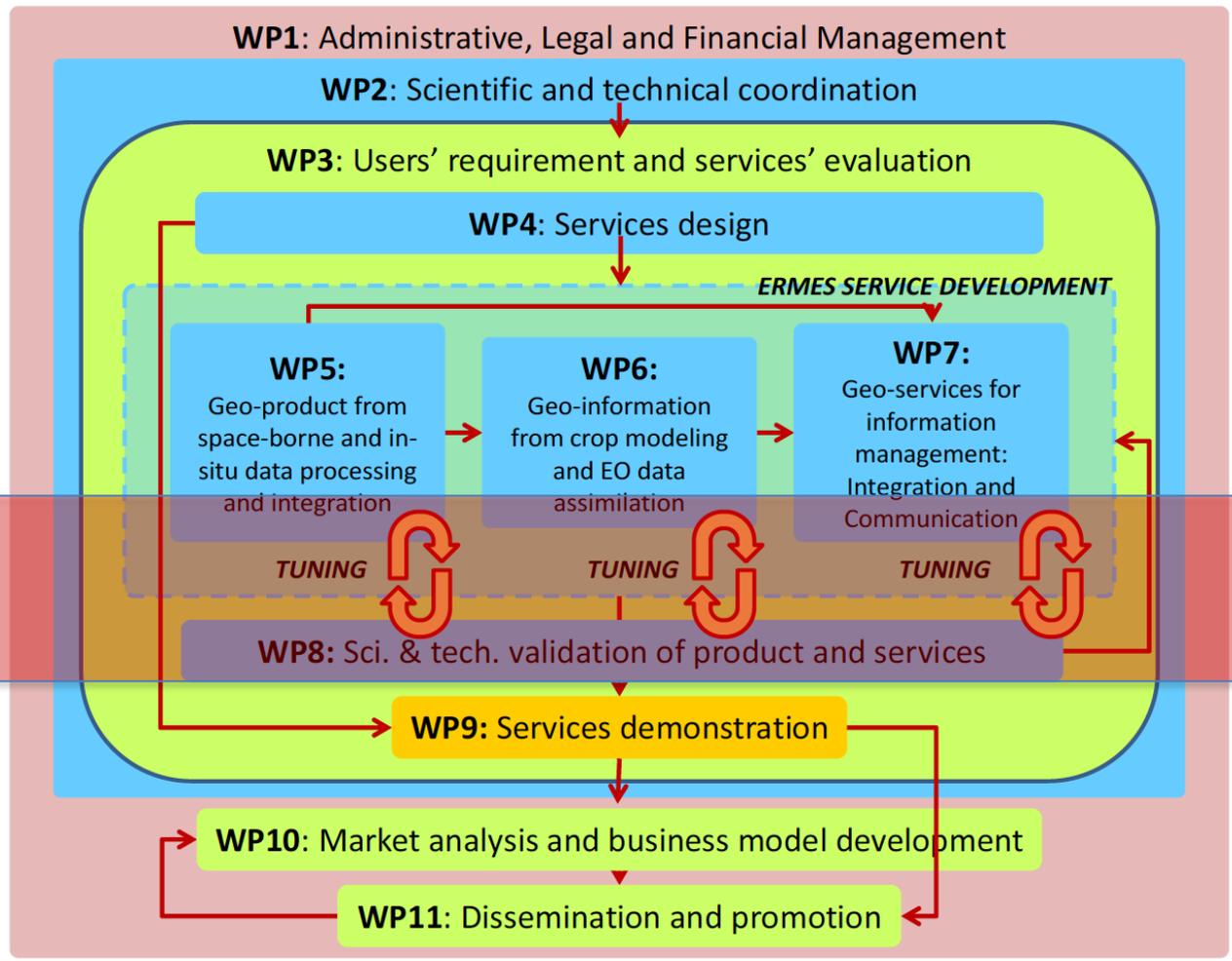
- ✓ **Introduction: WP Objectives and workplan**
- ✓ Activities
- ✓ Results
- ✓ Final Remarks

Main Objectives of the WP8

Assess **products** and **services** quality:

- ✓ EO products
- ✓ crop model information
- ✓ the overall service functionalities

feedbacks



 scientific soundness
 operational needs (users)

Work Plan for Months 18-34

	Sep-15	Oct-15	Nov-15	Dec-15	Jan-16	Feb-16	Mar-16	Apr-16	May-16	Jun-16	Jul-16	Aug-16	Sep-16	Oct-16	Nov-16	Dec-16	Jan-17	Feb-17	
Deliv	M19	M20	M21	M22	M23	M24	M25	M26	M27	M28	M29	M30	M31	M32	M33	M34	M35	M36	
D8.1	ERMES local products technical and scientific validation: first year v0																		
D8.2								ERMES local products technical and scientific validation: second year v1											
D8.3	ERMES regional products technical and scientific validation: first year v0																		
D8.4								ERMES regional products technical and scientific validation: second year v1											
D8.5	ERMES service technical and scientific validation: first year v0																		
D8.6								ERMES service technical and scientific validation: second year v1											

Workplan of WP8 in months 18-34 with reference to expected deliverables

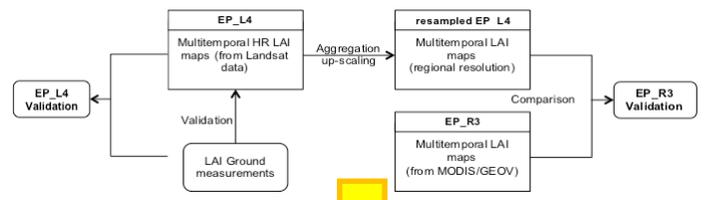
ERMES Validation: in a nutshell

Products, Info and tools

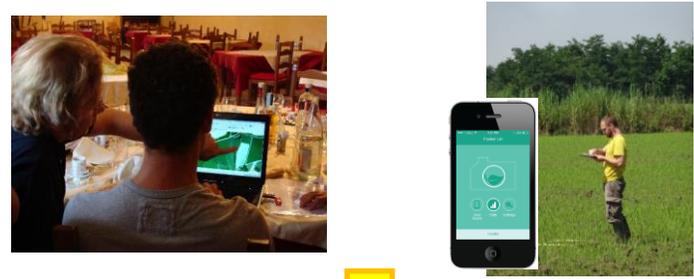
-  **Rice maps ; Flooding Maps** (EP_R1) *WP5*
-  **Phenology maps** (EP_R2) *WP5*
-  **Regional Blast Risk alert** (EI_R2) *WP6*
-  **Meteo data** (EP_R5) *WP5*
-  **LAI maps** (EP_R4 – EP_L4) *WP5*
-  **Yield forecast/estimation** (EI_R2) *WP5*
-  **Constant patterns** (EP_L2) *WP5*
-  **Seasonal patterns** (EP_L3) *WP5*
-  **Local Biotic risks; Development stage** (EI_Lx) *WP6*
-  **AgriNotebook smart app** *WP7*
-  **Regional and Local geoportals** *WP7*

Validation / Verification

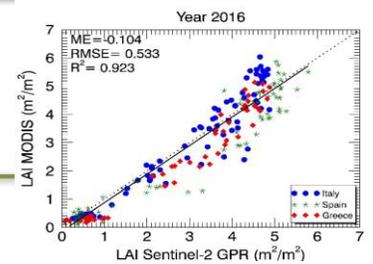
Protocols



Reference data /user feedbacks



Validation results

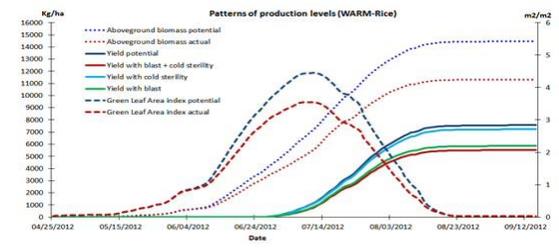


Tuning / Wrap-up

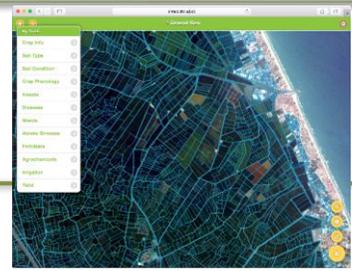
Processing chains



Modelling solutions



Tools



- ✓ Introduction: WP Objectives and workplan
- ✓ **Activities**
- ✓ Results
- ✓ Final Remarks

Activities months 18 - 34

First demonstration year

- finalize accuracy/quality assessment (single products/tools and of the overall services) for 2015
- analyse the lessons learnt after 2015 and propose tuning and improvements as regards processing chains, solutions and tools for the second year of demonstration

Second demonstration year

- accuracy/quality assessment for 2016
- analyse and recap the level of quality achieved by the products, tools and services at the end of the demonstration phase

Monitoring product quality throughout workflows and among countries

- ✓ Introduction: WP Objectives and workplan
- ✓ Activities
- ✓ **Results**
 - ✓ Lesson learned after first year of demonstration (2015 crop season)
 - ✓ Validation after second year of demonstration (2016 crop season)
- ✓ Final Remarks

from 2015 to 2016

Lesson learned 2015



Activities for/before 2016

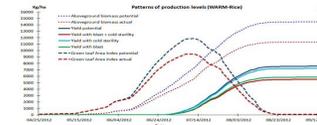
-  EP_R1: Rice mapping
-  EP_R2: Phenology
-  EP_R5 – EP_L5: Meteo data
-  EI_R2-4, EI_L1: Yield forecast/estimation
-  EI_R3- EI_L2: Risk alert
-  EP_R4-EP_L4: Bio parameter
-  EP_L2: Constant pattern
-  EP_L3: Seasonal pattern
-  EI_R1 & EP_L6 → Tools: Geoportals & smart app

recalibrate variables

new version validated before 2016



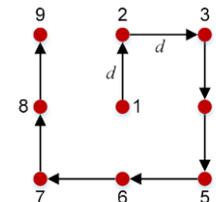
new regional solution



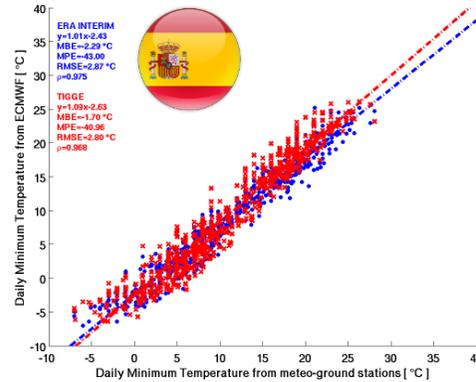
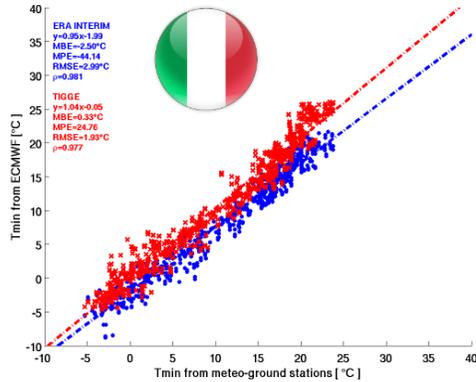
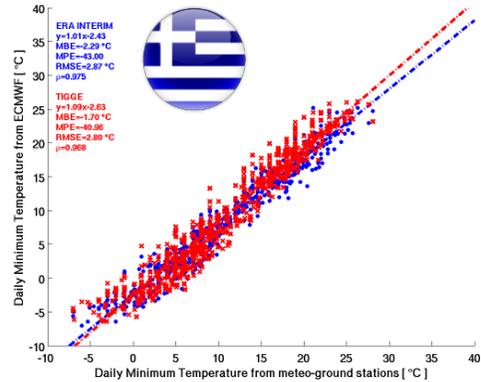
To improve LAI in situ data collection

LAI working group

improved data set 2016



meteo data re-processing



Recognised bias in V0

AREA	Statistical scores for TMax				Statistical scores for TMin			
	ρ	MPE [%]	MBE [°C]	RMSE [°C]	ρ	MPE [%]	MBE [°C]	RMSE [°C]
GR (no cal.)	0.98	14.64	-2.42	3.13	0.98	36.98	-3.48	3.81
GR (cal.)	0.98	1.08	-0.21	1.84	0.98	19.14	-1.86	2.39
IT (no cal.)	0.96	9.79	-2.22	3.46	0.96	34.31	-2.23	3.12
IT (cal.)	0.97	5.75	-0.07	2.30	0.96	7.45	0.16	2.15
ES (no cal.)	0.94	15.93	-3.34	4.02	0.95	23.421	-2.71	3.36
ES (cal.)	0.95	0.93	-0.18	2.03	0.96	6.14	-0.84	1.97

MPE [%]
36.98
19.14
34.31
7.45
23.421
6.14

intercalibration with MARS data sets (for archive and NRT) Improved performance in V1

Example → Archive (2003 – 2014)

PRESENTATION OUTLINE

- ✓ Introduction: WP Objectives and workplan
- ✓ Activities
- ✓ **Results**
 - ✓ lesson learned after first year of demonstration (2015 crop season)
 - ✓ **Validation after second year of demonstration (2016 crop season)**
 - A. EO product (EP_*)
 1. against reference data
 2. Indicators of internal consistency
 3. comparison with related variables
 4. intercomparison among products
 5. against expert knowledge and other ancillary data (judgement)
 - B. Modelling information (EI_*)
- ✓ Final Remarks



ERMES



EO product (EP_*)

1) Comparison against reference data

Rice crop maps (EP_R1)

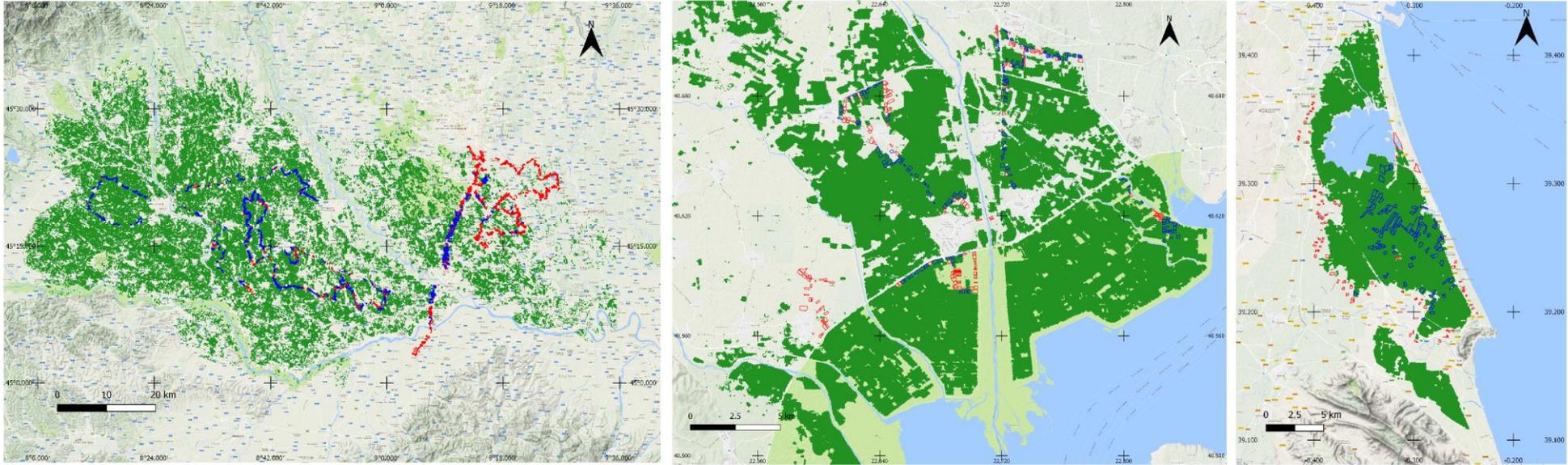


Table S2. Confusion matrices related to rice crop maps.

Italy

	Classification RICE	Classification NOT RICE	Producer's Accuracy
Reference RICE	1639	194	89,4
Reference NOT RICE	157	1237	88,7
User's accuracy	91,3	86,4	

Greece

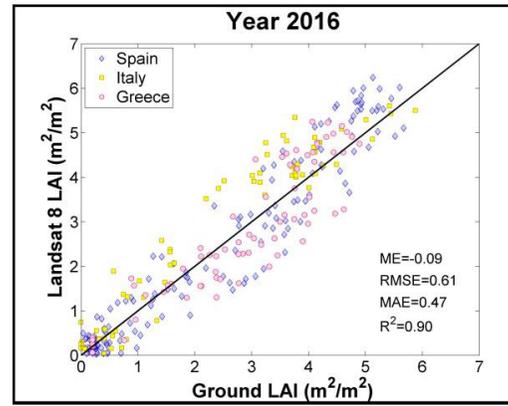
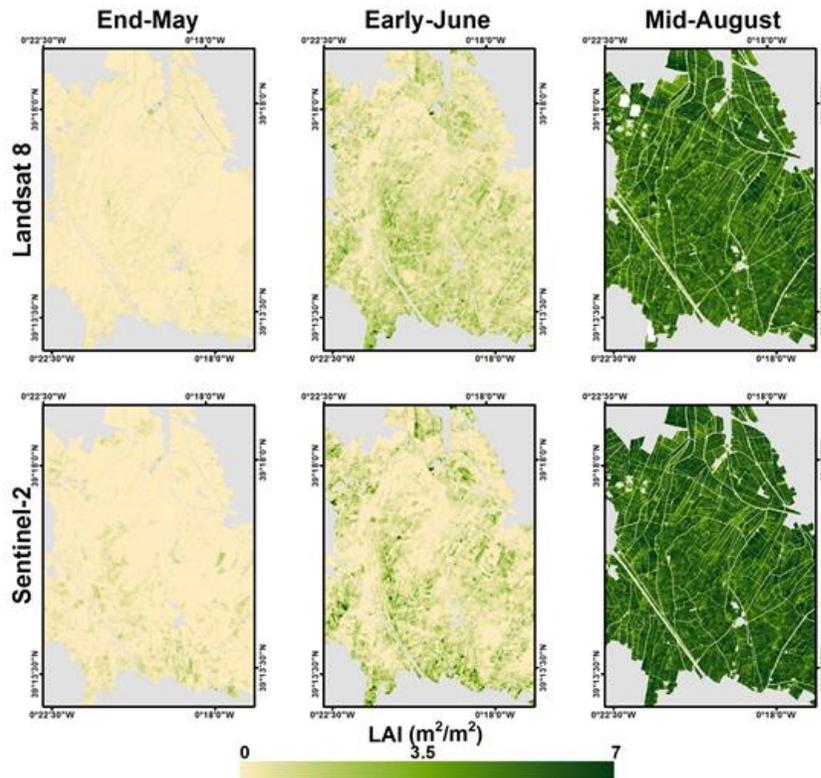
	Classification RICE	Classification NOT RICE	Producer's Accuracy
Reference RICE	7812	369	95,5
Reference NOT RICE	143	7364	98,1
User's accuracy	98,2	95,2	

Spain

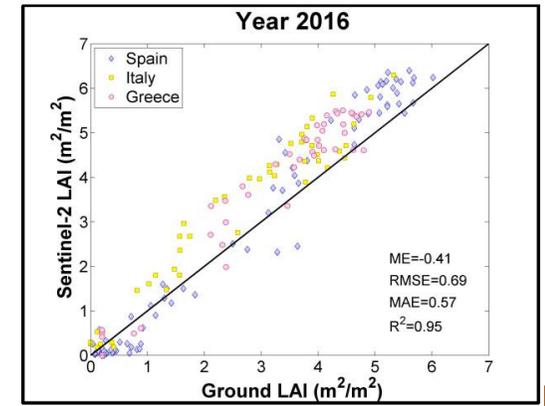
	Classification RICE	Classification NOT RICE	Producer's Accuracy
Reference RICE	21606	186	99,1
Reference NOT RICE	9	7367	99,9
User's accuracy	100,0	97,5	

1) Comparison against reference data

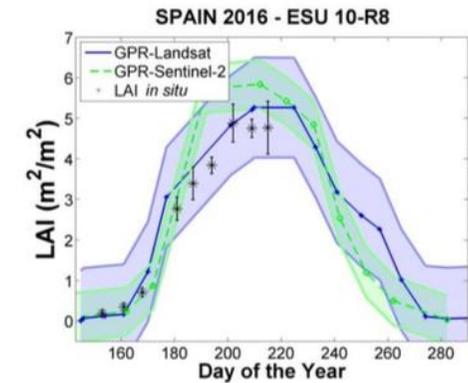
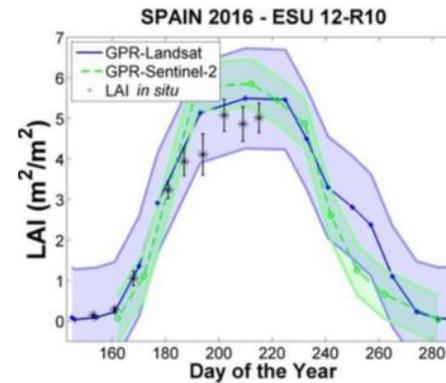
High Resolution LAI maps vs field data (EP_L4)



a)



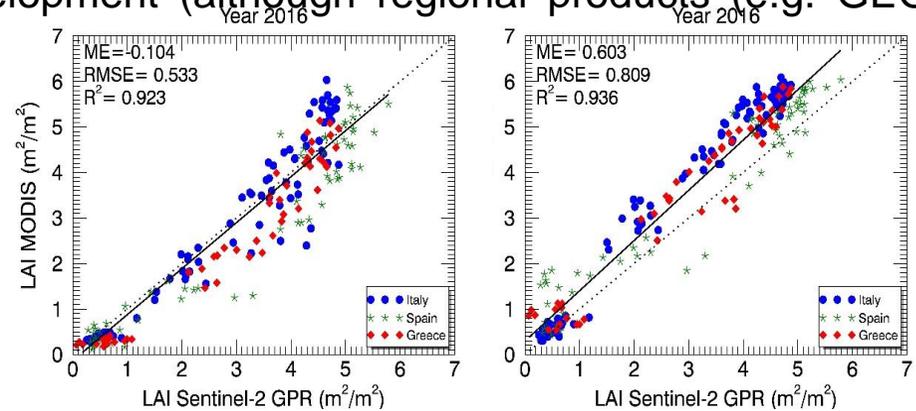
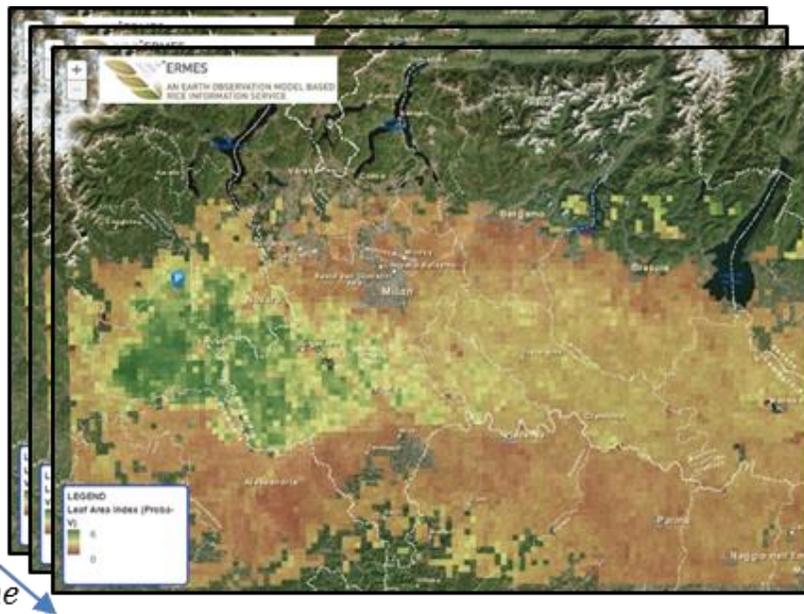
b)



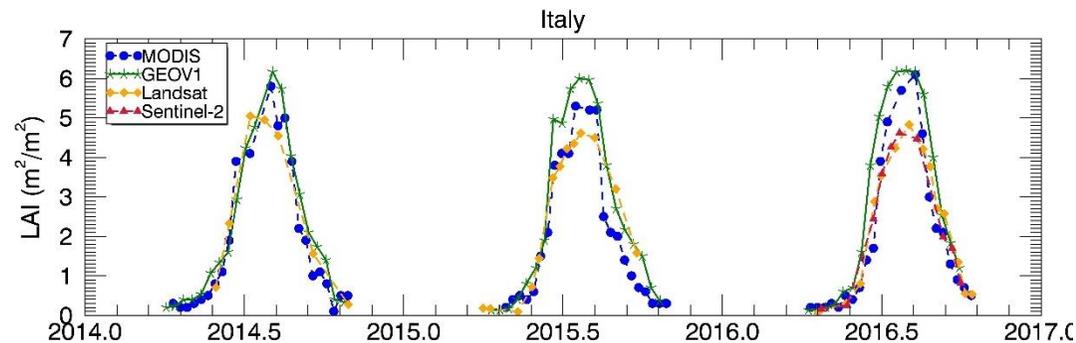
1) Comparison against reference data

Copernicus (EP_R3) vs HR LAI maps (EP_L4)

- ✓ A good consistency (high correlation, low bias) between the regional and local aggregated LAI maps
- ✓ similar dynamics from sowing up to rice development (although regional products (e.g. GEOV1) present higher values during the peak season)



Comparison regional and aggregated local LAI for the 3 ERMES study area during year 2016



Temporal evolutions in the 2014, 2015 and 2016 rice seasons for a representative rice pixel in Italy.

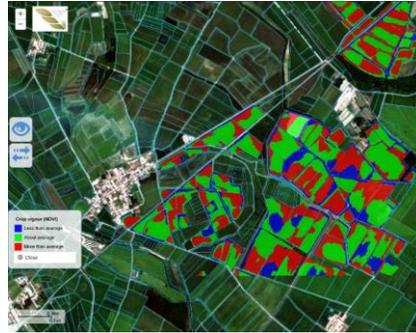
2) Internal validation by validity measures

Computation of “silhouette index” (cluster optimality): constant patterns (EP_L2)

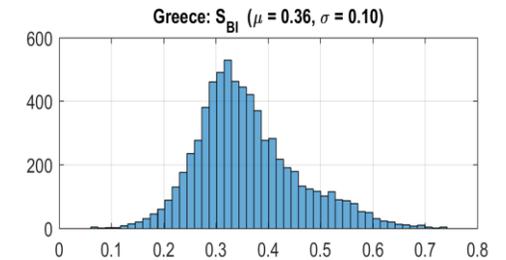
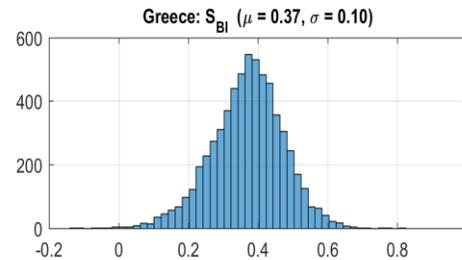
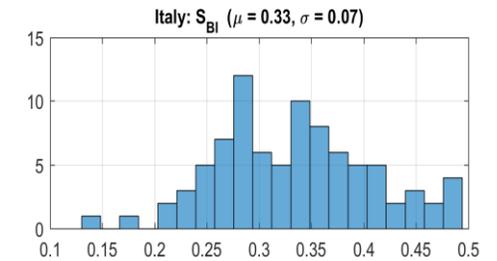
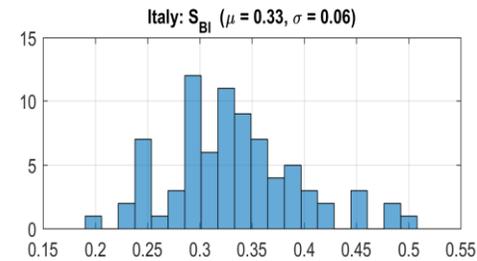
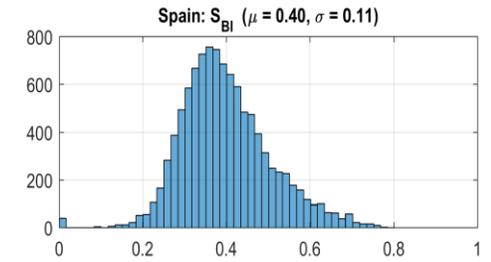
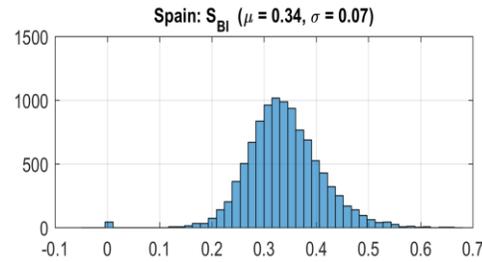
Soil Quality



Crop Vigour



Good separability of clusters



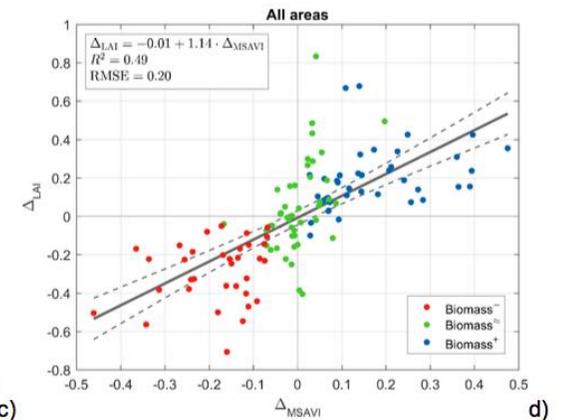
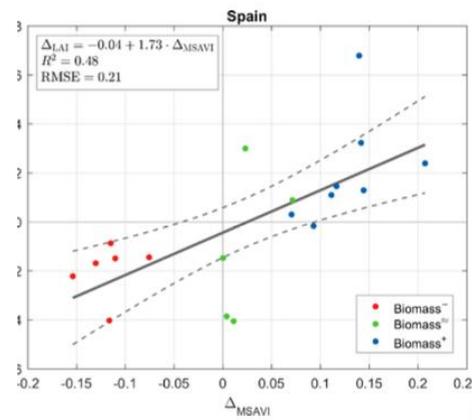
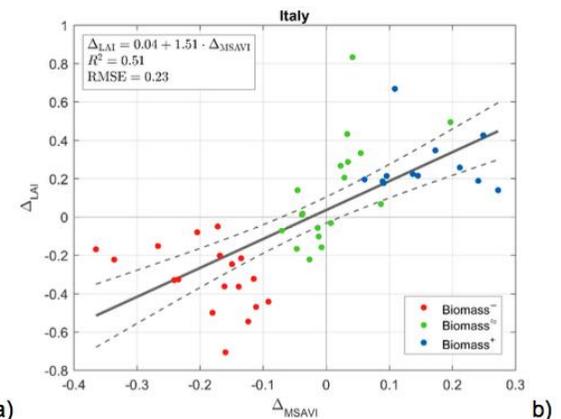
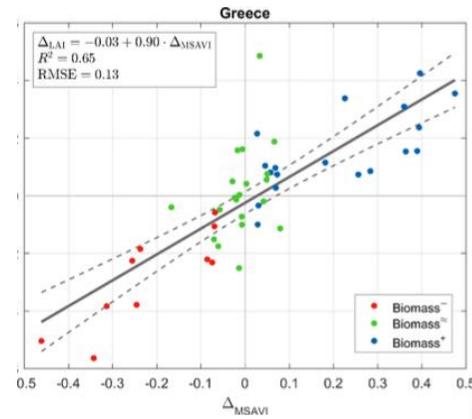
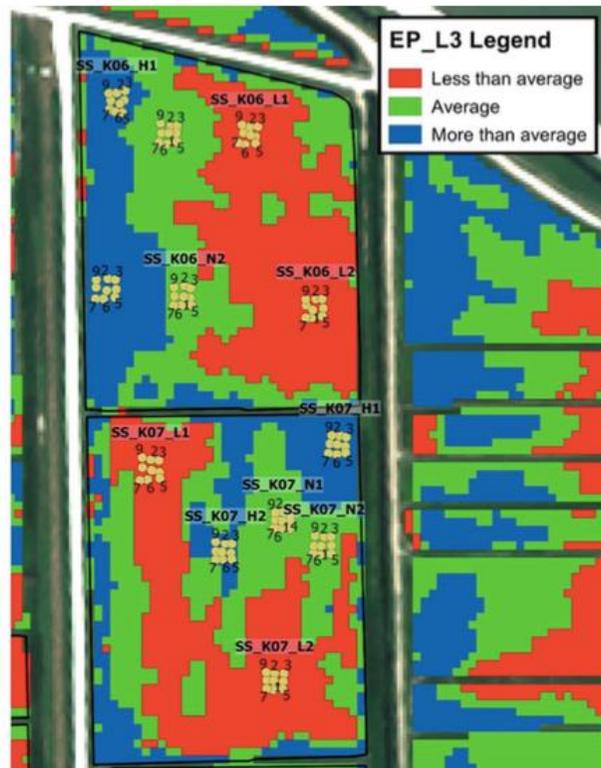
Intersection

$$s_j = \frac{b_{pj} - a_{pj}}{\max\{a_{pj}, b_{pj}\}}$$

Euclidean distances of object (pixel) \mathbf{x}_j from the nearest and second nearest prototype (cluster centre).

3) Comparison with related parameters:

Coherence of MSAVI patterns with infield LAI anomalies



4) Intercomparison among products 1/2

Coherence of information provided by local spatial variability products (pairwise)

Products compared	Objective	Method	Results	Comment
EP_L2 & EP_L3	constant patterns (EP_L2) tend to affect the current growing season's within-parcel variability (EP_L3) ?	one-way ANOVA and Tukey's HSD test EP_L3 variability maps vs EP_L2 for the second part of the season	significant differences in mean Δ_{MSAVI} values for all EP_L2 classes	a substantial number of parcel presented seasonal variabilities during the 2016 growing season that were also identified in previous years
EP_L3 & EP_L4	within-parcel biomass variability (EP_L3) explains the variability in LAI (EP_L4) during the growing season?	Correlation analysis EP_L3 variability maps (Δ_{MSAVI}) and a corresponding LAI variability measure Δ_{LAI} from the EP_L4 LAI maps	it confirms the correlation between the EP_L3 variability maps and the within-parcel variability of LAI	this confirms the suitability of the EP_L3 for supporting VRT surface fertilisation activities
EP_L3 & EP_L7	biomass variability (EP_L3) and crop homogeneity (EP_L7) at different stages of the development cycle do correspond?	one-way ANOVA and Tukey's HSD test EP_L3 variability maps (Δ_{MSAVI}) against the EP_L7 single-date classifications (before and after the booting stage)	significant differences between the Δ_{MSAVI} mean values for each EP_L7 class	distributions with high degree of overlap, some complementarity of products

Outcomes:

- Coherence → constant/seasonal patterns; seasonal patterns/LAI maps
- Complementarity → EP_L7 shows variabilities that are not detected by other products

4) Intercomparison among products 2/2

Example: intercomparison between products: EP_L3 vs EP_L7

Coherent as development is favourable or faster, opposite something is affecting the biomass

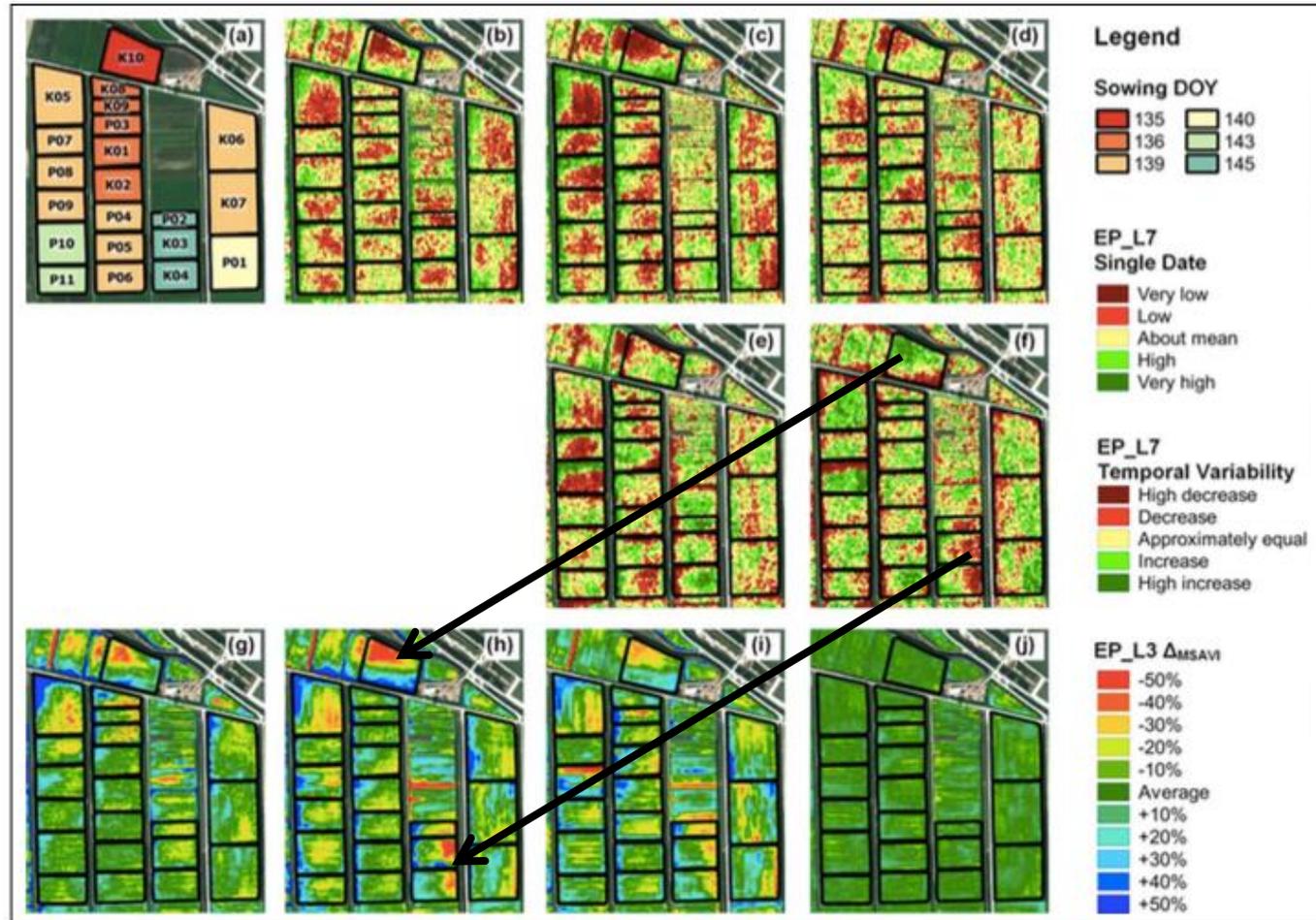
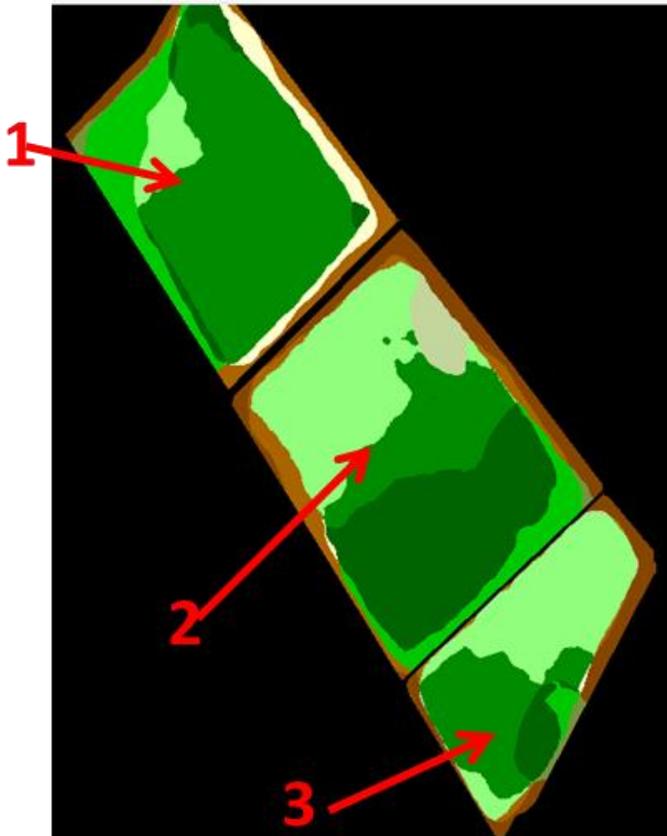


Figure 6.2: Various ERMES products produced during the 2016 growing season for some parcels inside DEMETER's stations in the Greek local study area. a) Sowing DOYs and names the parcels mentioned in the main text; the single-date EP_L7 product at DOY: b) 159, c) 167 and d) 175; the temporal variability EP_L7 product at DOYs: e) 159–167 and f) 167–175; the EP_L3 Δ_{MSAVI} products at DOY: g) 162, h) 174, i) 184 and j) 211.

5) Against expert knowledge and other ancillary data 1/2

Validation against expert knowledge: constant pattern (EP_L2)

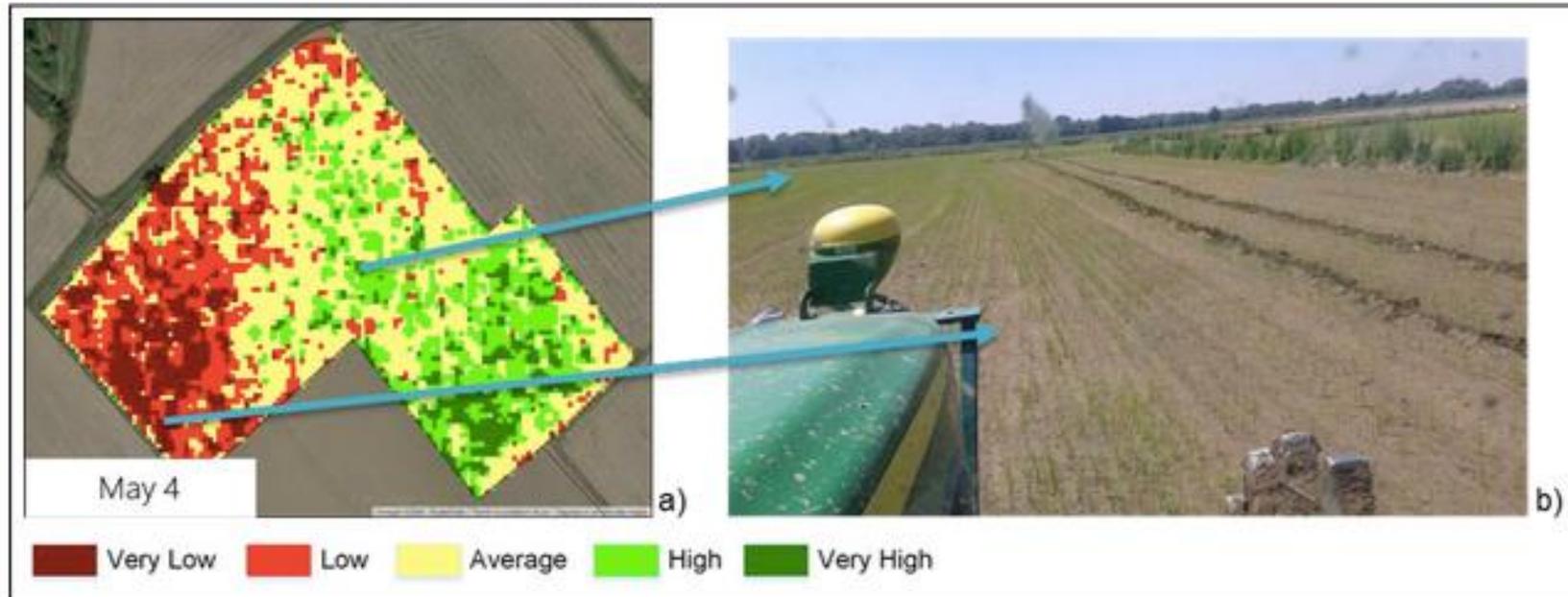


The constant patterns retrieved by EP_L2 map is explained by farmer knowledge:

- Zone 1 is to be ascribed to the water provision from this upper left area (red arrow).
- Zone 2 crop vigour differences in the map are related to a different soil texture.
- Zone 3 is almost due to the fact that this zone is interested by a overlapping by fertilizer spreader.

5) Against expert knowledge and other ancillary data 2/2

Validation against expert knowledge: early stages homogeneity maps (EP_L7)



Anomaly in SAR product (provided in NRT) reveals a strong correspondence to poor emergence in the field



■ ERMES

Modelling information (EI_{*})

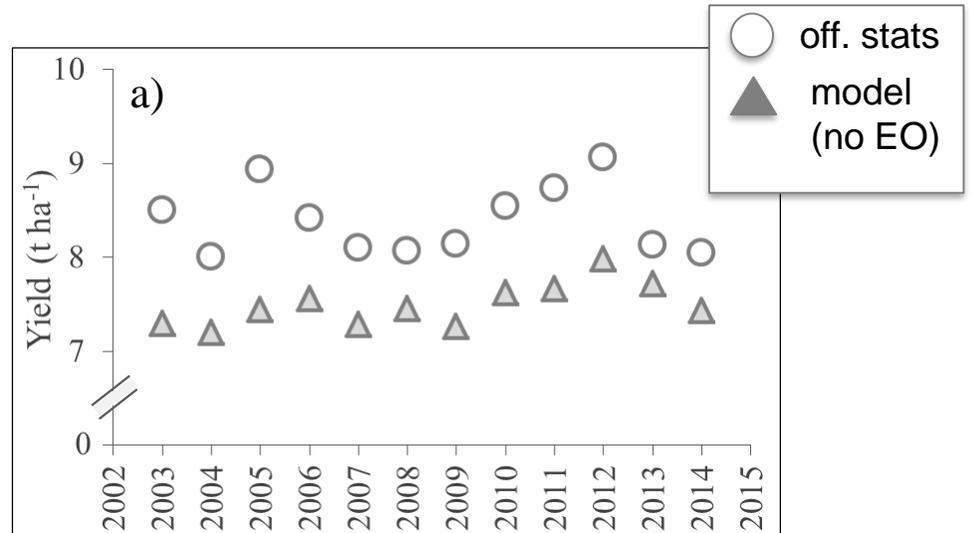
Regional Modelling solution (EI_R*)

Yield estimation against official statistics

Comparison between official and yields forecasted for Japonica cultivars in Valencia using

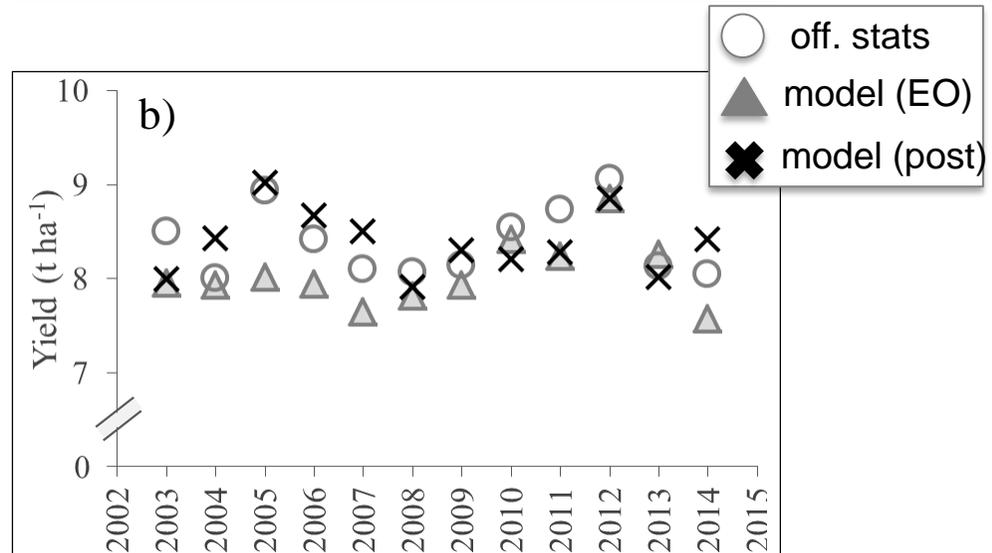
a) no EO products

- only the crop model

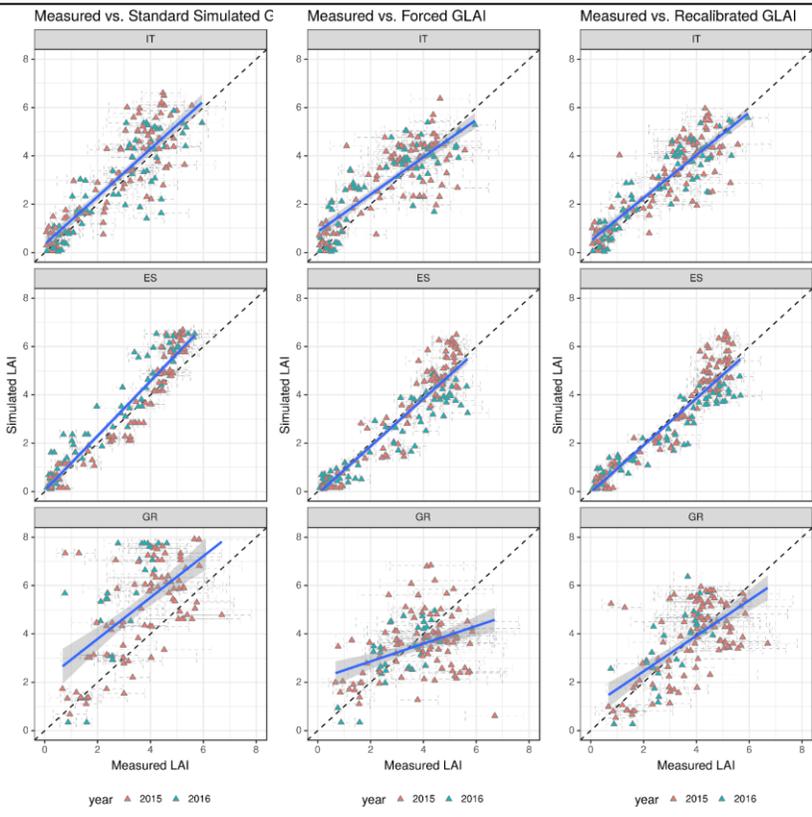


b) with EO products assimilation

- Only assimilation of remote sensing LAI
- statistical post-processing of simulated results (including LAI assimilation).



Simulated LAI against in situ data (assimilation of EO data)



Country	Model	2016						
		N	Slope	Int.	MAE	RMSE	EF	R ²
Italy	Default	105	0.95	0.25	0.68	0.77	0.79	0.76*
	Recalibrated		0.93	0.34	0.51	0.56	0.89	0.86*
Spain	Default	167	1.13	0.36	0.75	0.62	0.89	0.93*
	Recalibrated		0.77	0.19	0.58	0.47	0.93	0.92*
Greece	Default	33	1.64	0.71	2.82	2.97	-6.93	0.59*
	Recalibrated		1.12	-0.01	0.85	0.99	0.11	0.58*

Yield estimations at parcel level

- original model exhibited a generally satisfactory accuracy (average absolute errors of 1.69 t/ha and 0.77 t/ha obtained in the 2016 datasets for Greece and Italy, respectively).
- incorporation of EO information increased the model's accuracy (**improvement of approximately 23-30%**)

WP8 – FINAL REMARKS

- ✓ Big effort in
 - ✓ collecting in situ data (using homogeneous protocol) and maximize their usefulness
 - ✓ interacting with users (single product and service levels)
- ✓ quality of EO products @different scales proved to be adequate
- ✓ importance of EO data assimilation into modelling solutions
- ✓ suitability of tools for providing value added information as usable and in time



In situ data



Satellite data



WARM model



Smart app.



Geoportal

Question Time