

MedCOF Mediterranean Climate Outlook Forum

01/06/2022

Climate services

Prototypes for agriculture and forestry sectors

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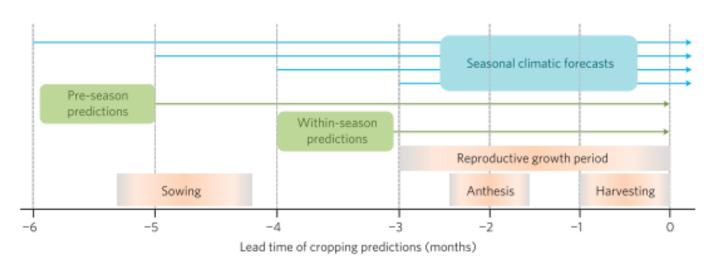


Figure 1 | Timing of cropping predictions. The cropping calendar illustrates the times at which the pre- and within-season predictions of crop failures and yield levels were conducted and the lead times of seasonal climatic forecasts on a monthly basis.



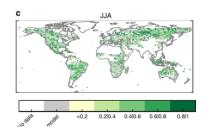
Prediction of seasonal climate-induced variations in global food production

Toshichika lizumi¹*, Hirofumi Sakuma^{2,3}, Masayuki Yokozawa¹, Jing-Jia Luo⁴, Andrew J. Challinor^{5,6}, Molly E. Brown⁷, Gen Sakurai¹ and Toshio Yamagata³

Skilful forecasting of global fire activity using seasonal climate predictions

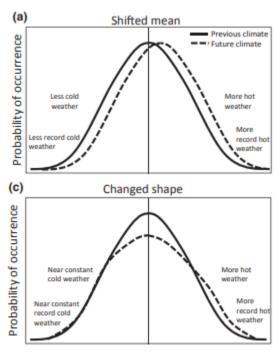
Marco Turco 1, Sonia Jerez², Francisco J. Doblas-Reyes³,4, Amir AghaKouchak 5, Maria Carmen Llasat 1, & Antonello Provenzale6

NATURE COMMUNICATIONS | (2018)98278 | DOI: 10.1038/s41467-018-05250-0 | www.nuture.com/instrusecommunications

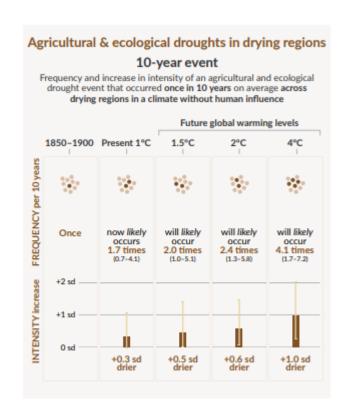








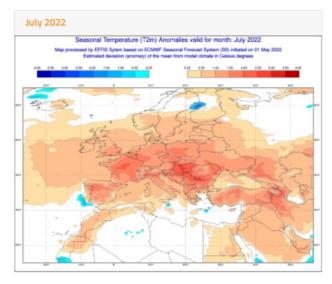
Thornton et al, 2014

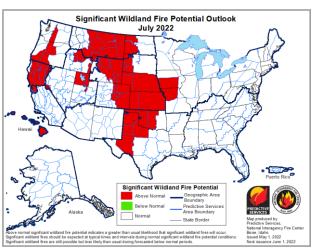


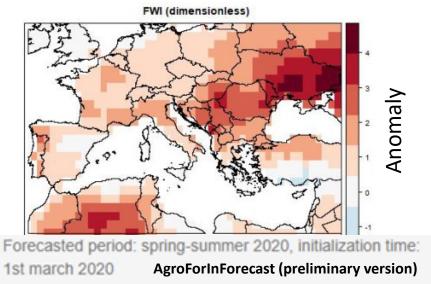
IPCC sixth report, 2021

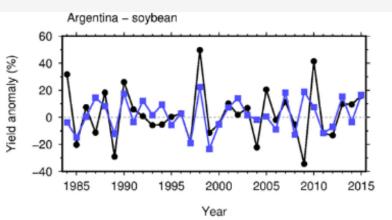












T. Iizumi et al. Climate Services 11 (2018) 13-23





J. Bedia et al./Climate Services 9 (2018) 101–110

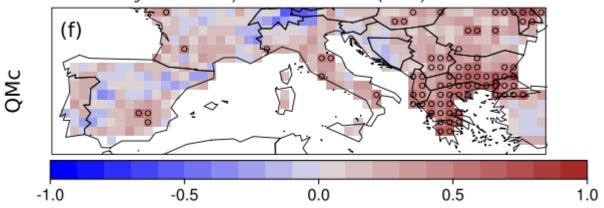


Fig. 3. ROC Skill Score of the System4 FWI predictions

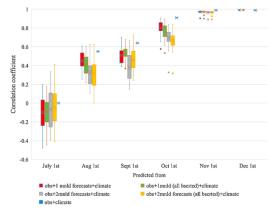


Fig. 4. Correlation coefficient between simulated yield using CFSv2 forecasts (all members) and yield simulated using weather station data from Janakpur

P.K. Jha et al. Agricultural and Forest Meteorology 265 (2019) 349-358

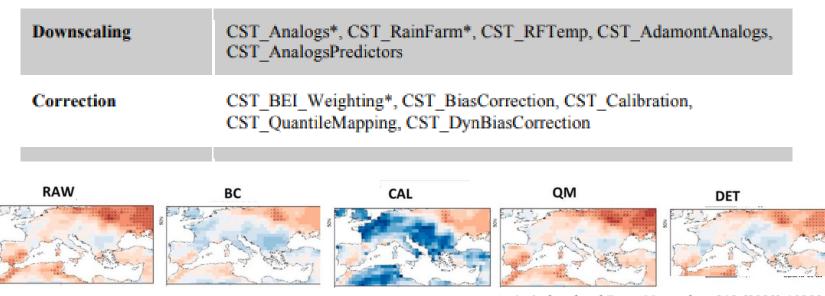






The CSTools (v4.0) Toolbox: from Climate Forecasts to Climate Forecast Information

Núria Pérez-Zanón ¹, Louis-Philippe Caron ^{1,2}, Silvia Terzago ³, Bert Van Schaeybroeck ⁴, Llorenç Lledó ¹, Nicolau Manubens ¹, Emmanuel Roulin ⁴, M. Carmen Alvarez-Castro ⁵, Lauriane Batté ⁶, Carlos Delgado-Torres ¹, Marta Domínguez ⁷, Jost von Hardenberg ^{8,3}, Eroteida Sánchez-García ⁷, Verónica Torralba ¹, Deborah Verfaillie ⁹

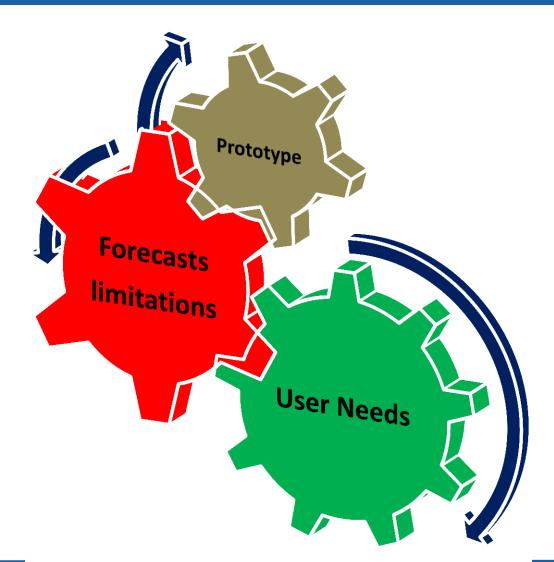


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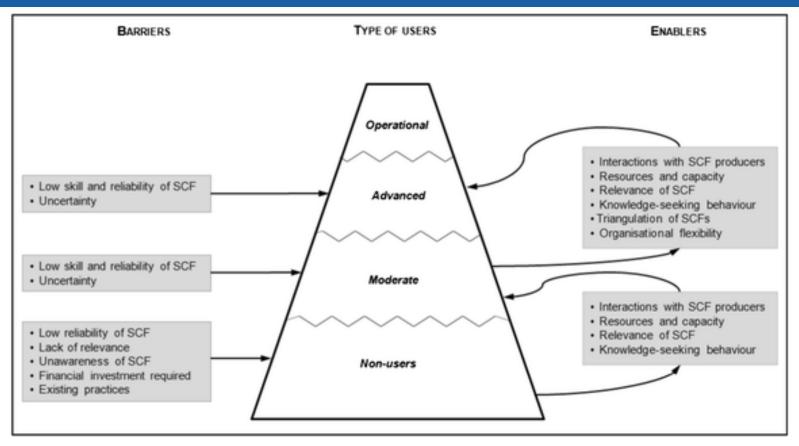


Co-Design









Climatic Change (2016) 137:89–103 DOI 10.1007/s10584-016-1671-8

Barriers and enablers to the use of seasonal climate forecasts amongst organisations in Europe

Marta Bruno Soares1 · Suraje Dessai1





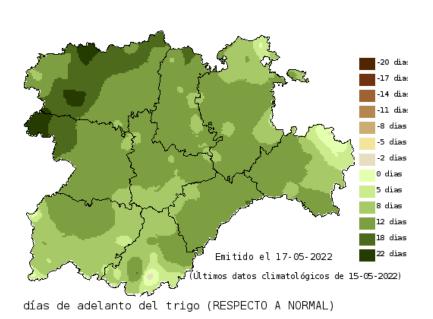


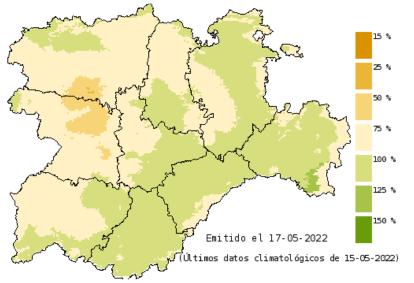


http://cosechas.itacyl.es/

Seasonal forecasts for estimation of cereal yield in Castilla y León







% trigo esperado respecto al promedio de 30 años



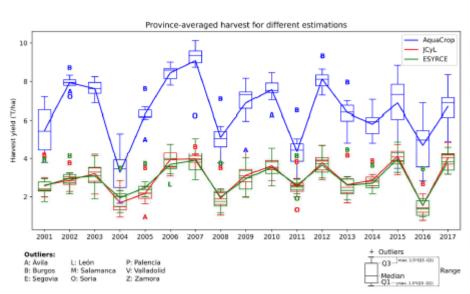


Figure 12: Time series of cereal harvest over Castilla y León region coming from Aquacrop reference experiment (blue), ESYRCE (green) and JCYL (red) databases. Boxplots represent the spread between different values at province level.

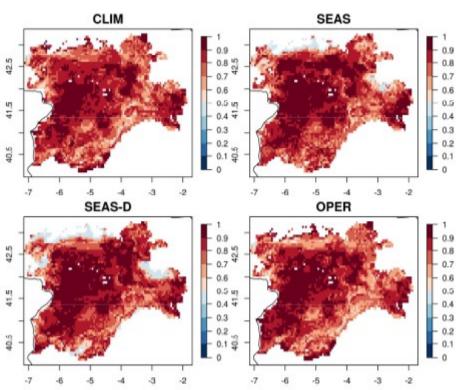


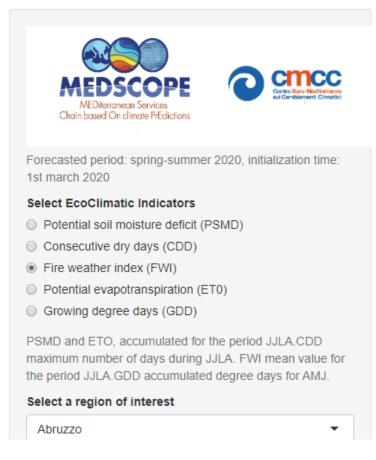
Figure 6: Maps of ROC area for lower and upper wheat yield tercile forecasts (reference simulations have been used as observations) over 1995-2018 hindcast period for the different experiments: CLIM (left-top), SEAS (right-top), SEAS-D (left-bottom) and OPER (right-bottom)

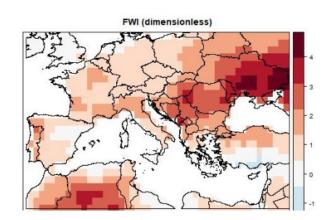
Memory effect on system inertia

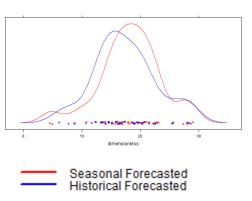


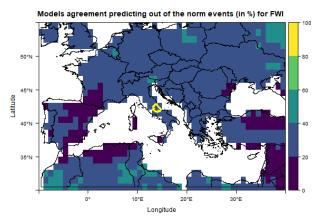


TESTING THE PROTOTYPE: Seasonal Forecasting of EcoClimatic Indicators









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Forecast accurracy for Abruzzo is 0.09 (from -1 to 1)





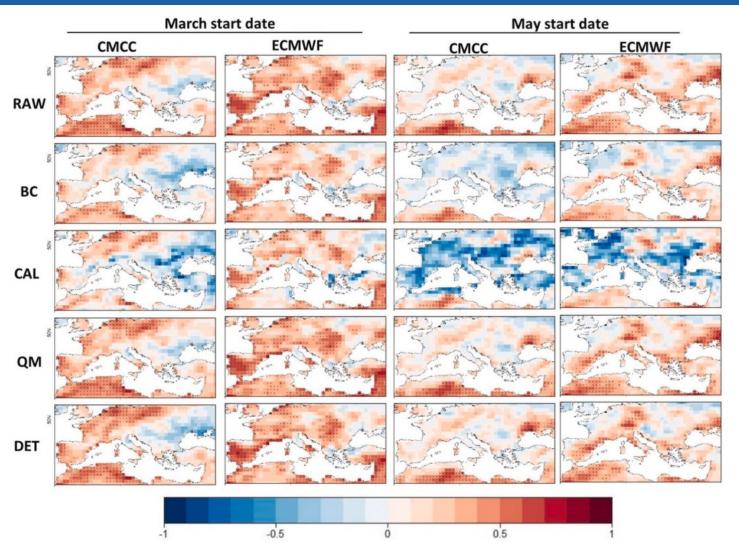


Fig. 4. Time series correlations (1993-2015) between the hindcast ensemble mean and observations (ERA5 reanalysis) for Potential Soil Moisture Deficit (PSMD) using ACC metric. Checked cells indicate a significant correlation (p-value < 0.05).





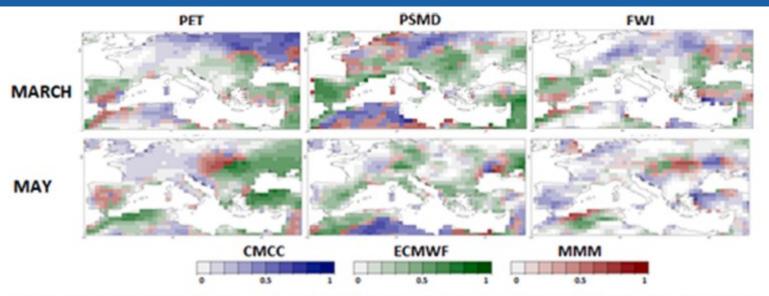


Fig. 6. Anomaly correlation coefficient (ACC) between SPSs and ERA5 for both forecast start dates after QM corrections. The figure shows the greatest ACC among mo

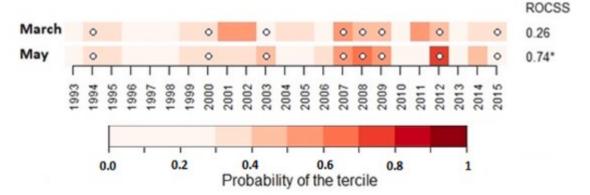


Fig. 8. Forecast skill of PET (ECMWF SPS and QM) for a selected area in central-eastern Europe (see red rectangle in Fig.3) using both start dates. Colours represent the probability of the upper tercile according to the hindcast, whereas white dots represent the observed tercile.









Figure 1: The Prométhée area



Figure 2: The Crau plain





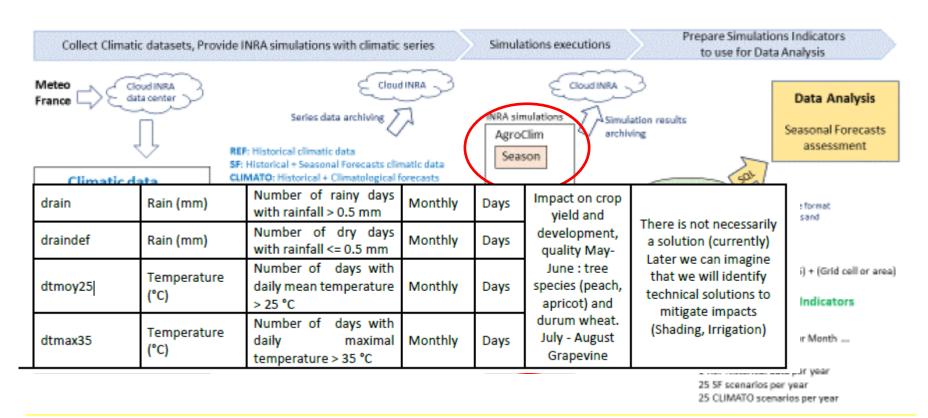


Figure 3: The MedScope INRAE workflow.



Prepare Simulations Indicators Collect Climatic datasets, Provide INRA simulations with climatic series Simulations executions to use for Data Analysis Carte des Sols Climat Scénarios climatiques Prévisions saisonnières & décennales fonctionnelle spatialisé Indicateurs Modélisation **Pratiques** Coussouls Prairies, Vergers, grandes (steppe) zone agricoles cultures humide. Grassland vield Dry cuts t/ha Optimization of Choice of the harvest Maximal and biomass per Grass C1, C2 minimal harvested (3 within date to optimize the grass, grass and C3 Temperature, the overall grass production averaged over the production Rainfall. surface of interest (grid) successive season) two Irrigatio production periods. Radiation, doses=f(so Wind, Relative Humidity Déveld Wheat fertlization, trade price, vield Maximal and dry biomass of 1 t/ha production per LAI Ser (Wheat) minimal harvested grain storage, price silo management season ppe Temperature, anticipation, Rainfall, fertlization Géométril Radiation,





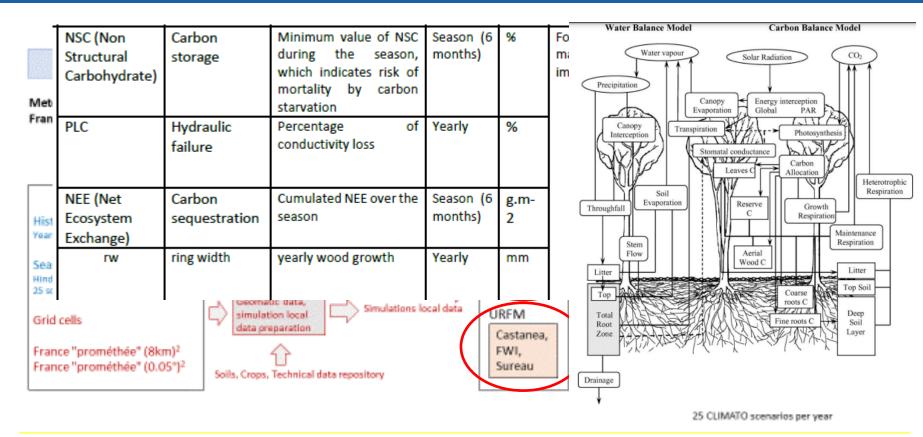


Figure 3: The MedScope INRAE workflow.





		AUC 1st T	AUC 3rd T	AUC 1st Q	AUC 5th Q	ACC
Grass	C C1	0.839	0.820	0.866	0.828	0.740
Grass	C2	0.541	0.589	0.574	0.637	0.155
Grass	C3	0.460	0.447	0.463	0.430	-0.070
Wheat		0.960	0.960	0.954	0.965	0.940



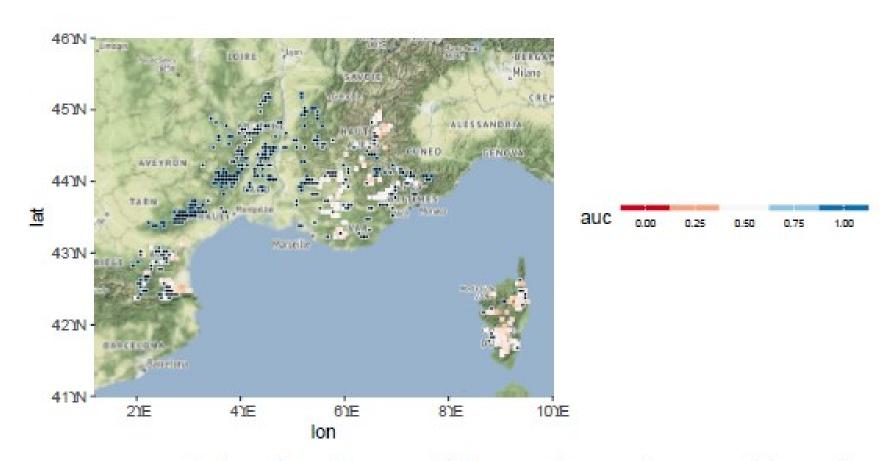


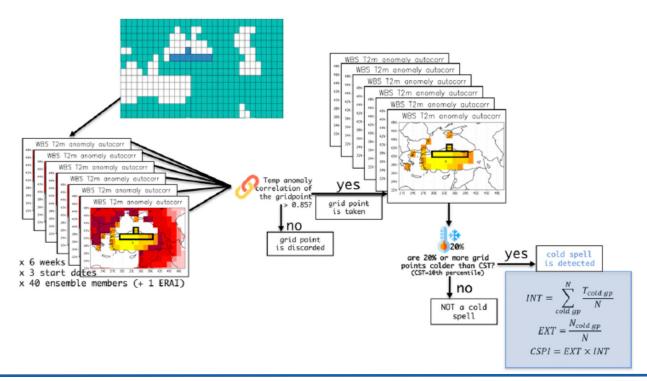
Figure 29: Spatial value of AUC (first tercile) for NSC of Fagus sylvatica with low Soil Water Capacity





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Concluding remarks

Different types of applications:

- Raw climate variables
- Complex indicators
- Process based models

Good practices:

- Assessing accuracy
- Assessing accuracy
- Assessing accuracy



Thank you for your attention

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