



# Yield prediction of durum wheat: the added value of MED-GOLD climate services products

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# Aim & scope

Early within-season weather conditions forecast and yield prediction can provide useful information to improve farmers' management decisions.



increase yield and reduce potential risk

The skills of the **ECMWF-System5 seasonal time-scale forecasting** provided through the *Copernicus Data Store* (CDS) were evaluated as a driver to the crop modelling system **DELPHI**



Case study:  
**durum wheat** yield prediction



Three hot spot areas in **Italy**

# The DELPHI model

The DELPHI integrated forecast system is a *mechanistic model*

INPUT DATA



- Durum wheat phenology
- Soil hydrology
- Soil and crop N content
- Roots growth

The model has been **calibrated and tuned for durum wheat and Italian conditions** since 1995

Europ. J. Agronomy 43 (2012) 108–118



Durum wheat modeling: The Delphi system, 11 years of observations in Italy

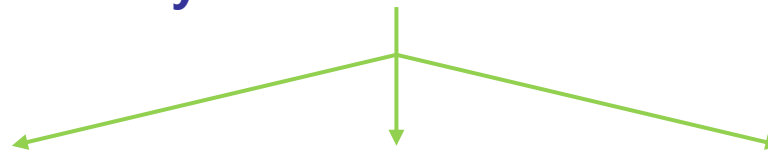
P. Toscano<sup>a,b,\*</sup>, R. Ranieri<sup>c,1</sup>, A. Matese<sup>a</sup>, F.P. Vaccari<sup>a</sup>, B. Gioli<sup>a</sup>, A. Zaldei<sup>a</sup>, M. Silvestri<sup>c</sup>, C. Ronchi<sup>c</sup>, P. La Cava<sup>c</sup>, J.R. Porter<sup>e</sup>, F. Miglietta<sup>a,d</sup>

The DELPHI model includes also the climatic reconstruction based on long-term observations in order to create **three synthetic weather scenarios**

**“DRY”**

**“AVERAGE”**

**“WET”**

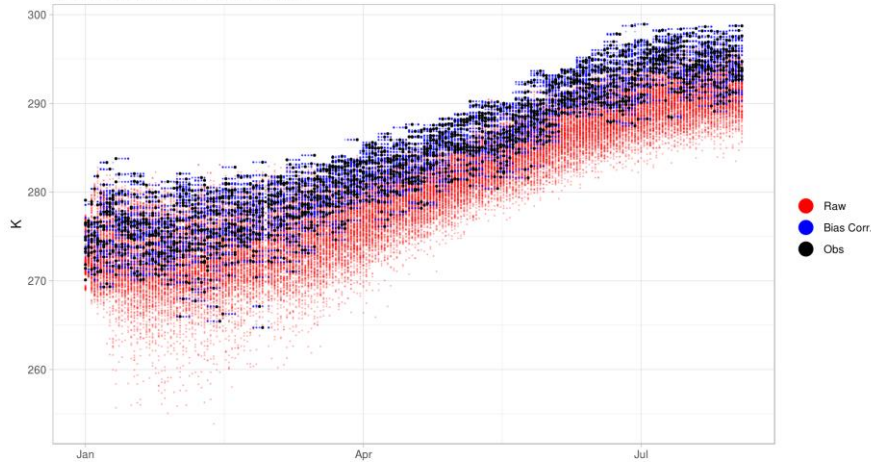


# Correction of biased dataset

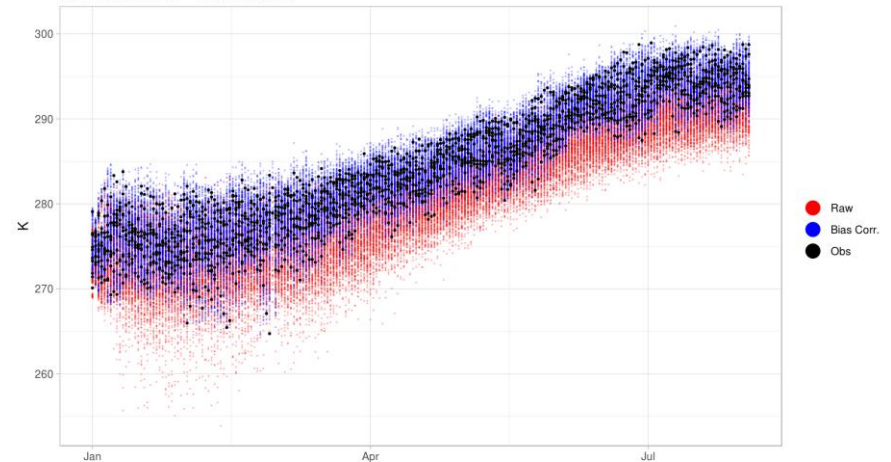
Biases in model simulation are commonly detected by *validation*.

As correction technique, we have performed a **quantile mapping methods**.

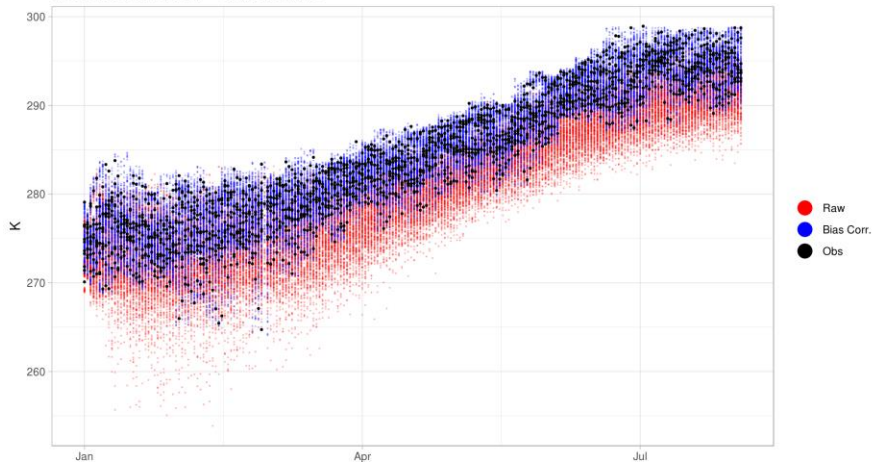
Forecast: ecmwf - Obs: ERA5 - Var: mn2t24  
Qmap method: QUANT - Sample length: 5



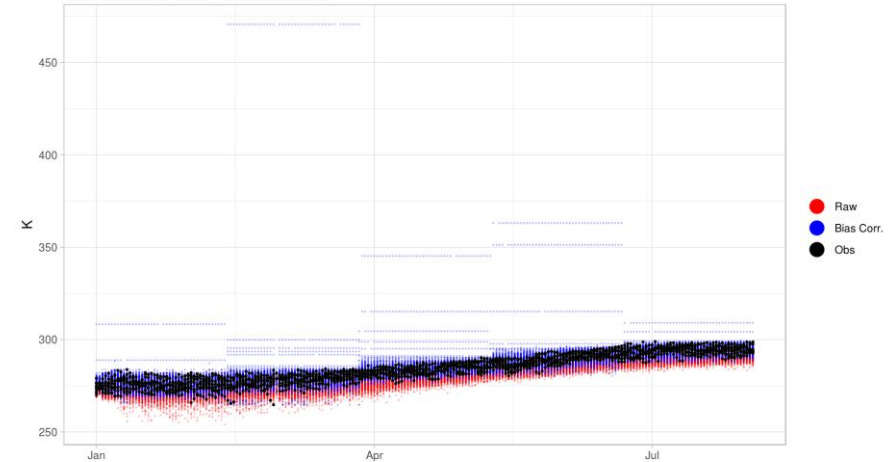
Forecast: ecmwf - Obs: ERA5 - Var: mn2t24  
Qmap method: PTF - Sample length: 5



Forecast: ecmwf - Obs: ERA5 - Var: mn2t24  
Qmap method: RQUANT - Sample length: 5



Forecast: ecmwf - Obs: ERA5 - Var: mn2t24  
Qmap method: SSPLIN - Sample length: 43



**Bias correction of temperature forecasts.** Each dot represents a daily value from raw seasonal forecast ensemble members (red), the reanalysis ERA5 (black) and the bias corrected daily values (blue).

# Crop season simulation benchmark ICT dataset

For both **unbiased and biased** weather forecast dataset, high and low performance crop year in terms of yield have been chosen:

- **BAD year** (below average yield): 2010 for Ravenna; 2007 for Foggia and Ancona
- **GOOD year** (above average yield): 2012 for Ravenna; 2016 for Foggia and Ancona

The DELPHI model was run with observed daily weather data from sowing to harvest



**REFERENCE YIELD**

Yield **hindcasts** were calculated at a *monthly time step*, starting from February 1st and April 1st, by feeding the DELPHI model with:

*New tool*

1. weather seasonal forecast (**25 ensemble** for 6 months of forecast) until the end of the growing season

vs

*Current mode*

2. synthetic weather scenarios based on historical observations (**dry**, **average**, **wet**)



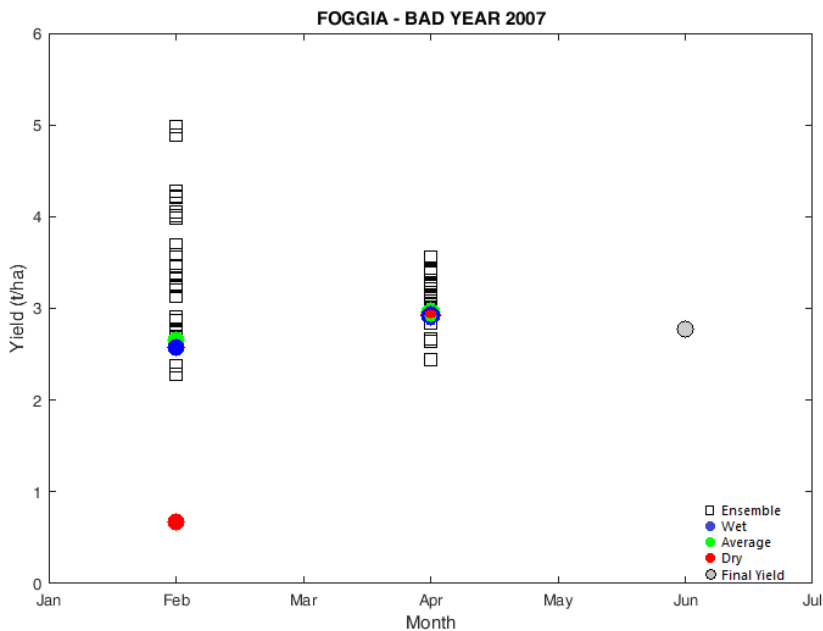
Observed Data Scenario Data

# Hindcast results

Yield prediction on the 1<sup>st</sup> June on the basis of three historical scenarios (red-green-blue circle) and on the basis of seasonal forecast (black square) are reported.

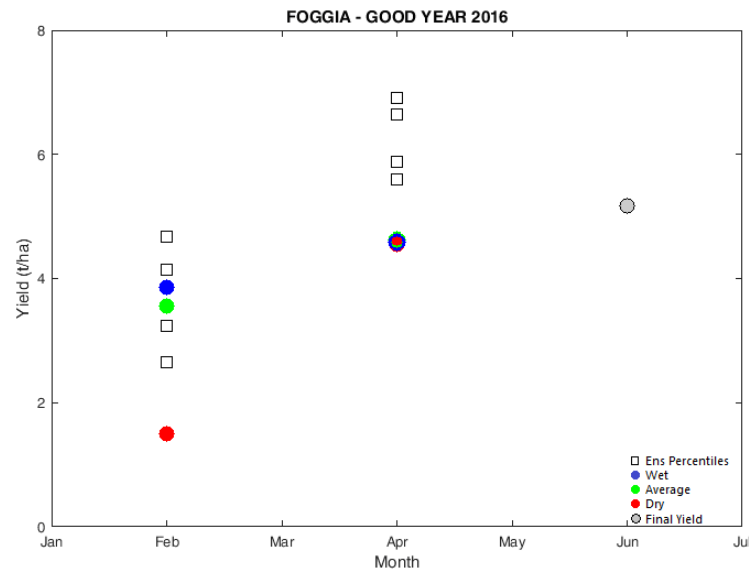
For each year the reference yield is also reported (grey circle).

## Biased dataset



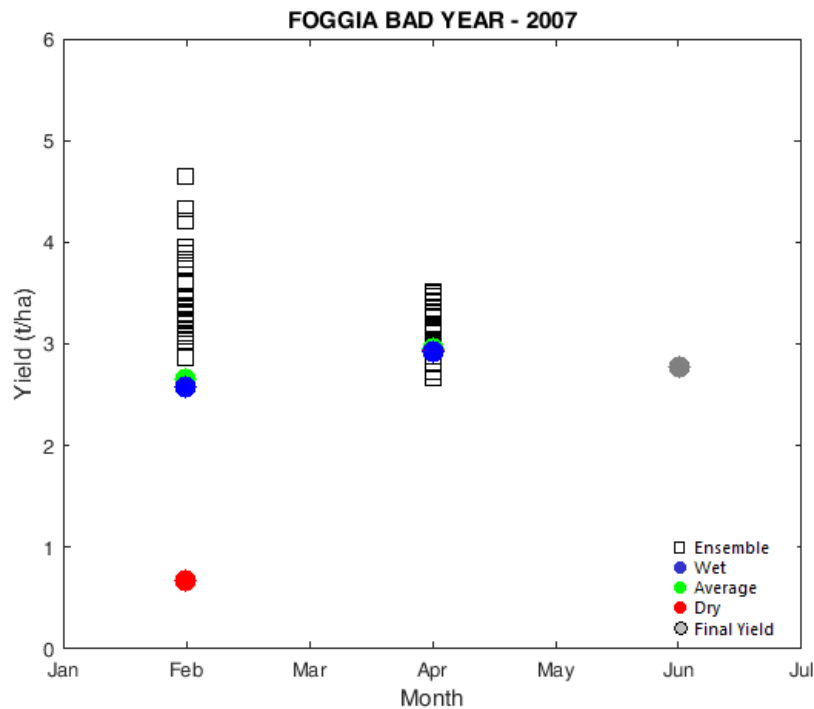
Seasonal forecast scenario range is narrower than that provided by the Dry-Average-Wet scenario on February 1st, wider on April 1st.

Seasonal forecast scenario range is narrower than that provided by the Dry-Average-Wet scenario on February 1st, wider on April 1st with a tendency to overestimate yield value.

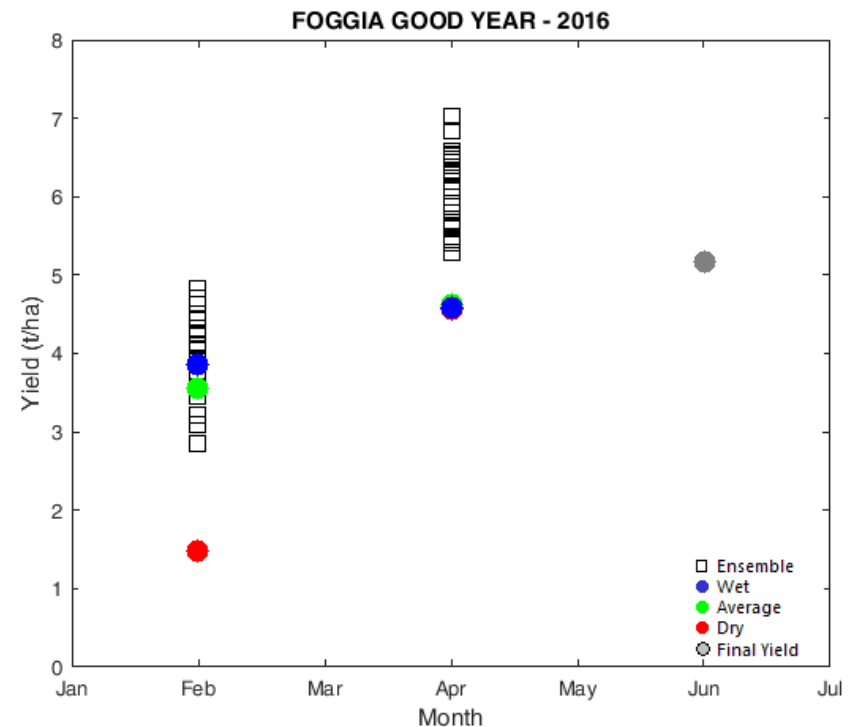


## Unbiased dataset

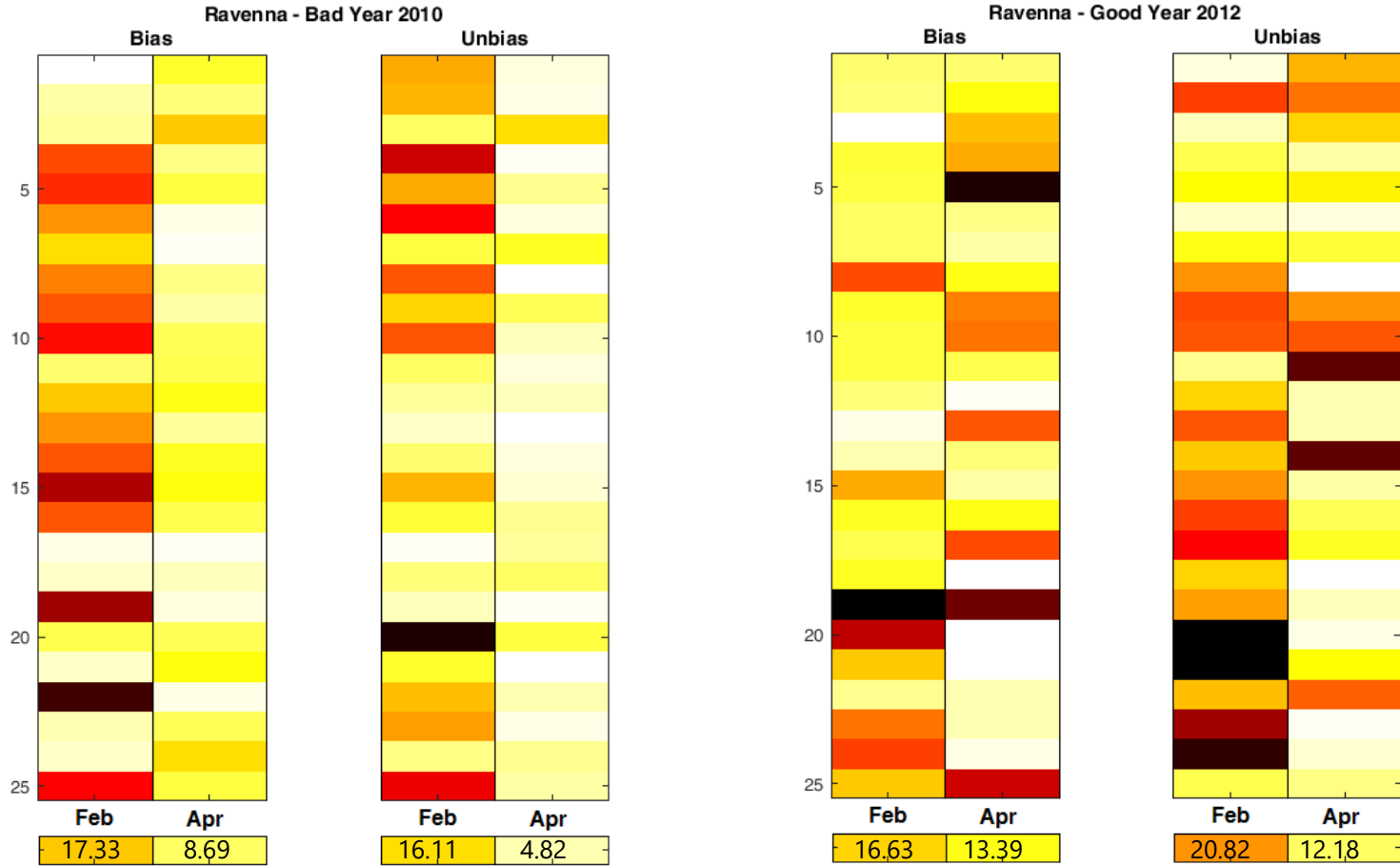
Seasonal forecast scenario range is wider in both dates than that provided by the Dry-Average-Wet scenario on February 1st, with a tendency to overestimate yield value.



Seasonal forecast scenario range is narrower than that provided by the Dry-Average-Wet scenario on February 1st, wider on April 1st with a tendency to overestimate yield value.

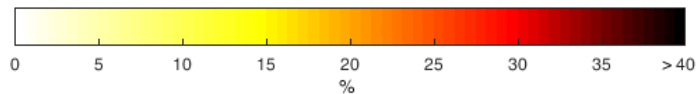


# Yield percentage variation between DELPHI *new tool* and *current mode* vs. reference value



	DRY	AVG	WET
<b>Feb</b>	1.10	9.50	3.24
<b>Apr</b>	7.75	18.93	25.44

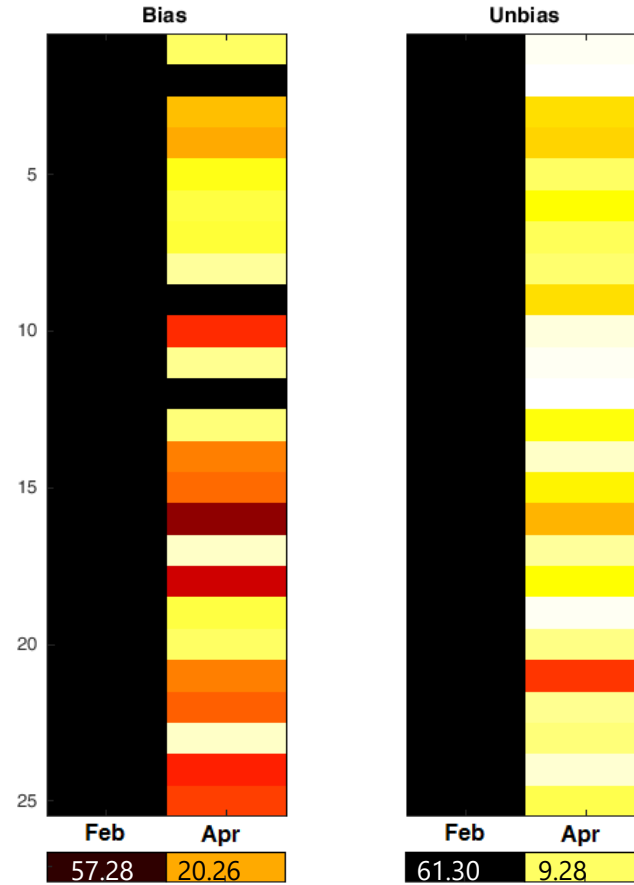
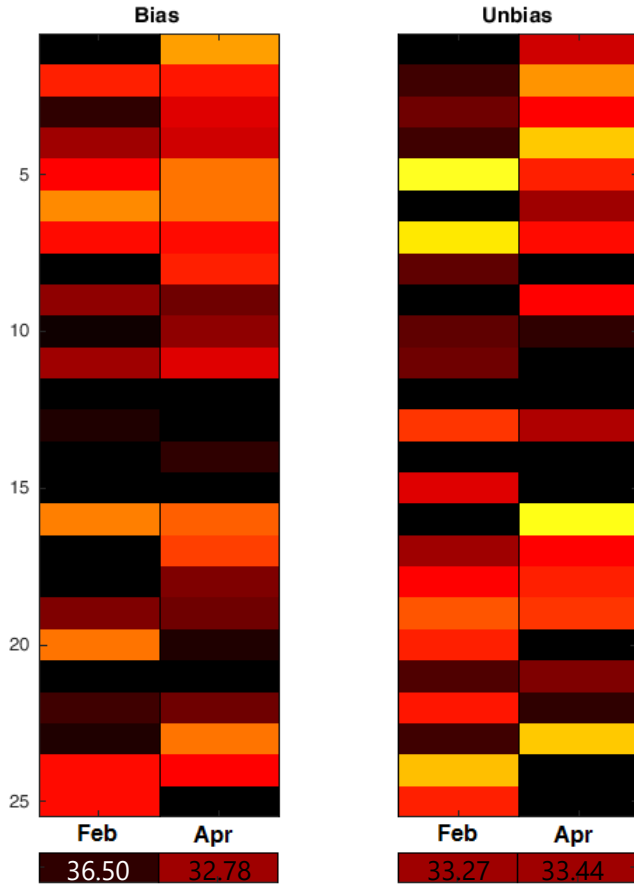
	DRY	AVG	WET
<b>Feb</b>	60.08	32.52	26.03
<b>Apr</b>	28.95	9.73	3.89





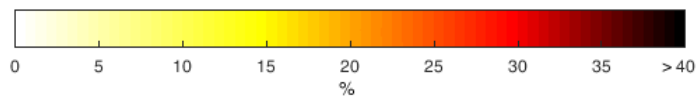
ANCONA - Bad Year 2007

ANCONA - Good Year 2016

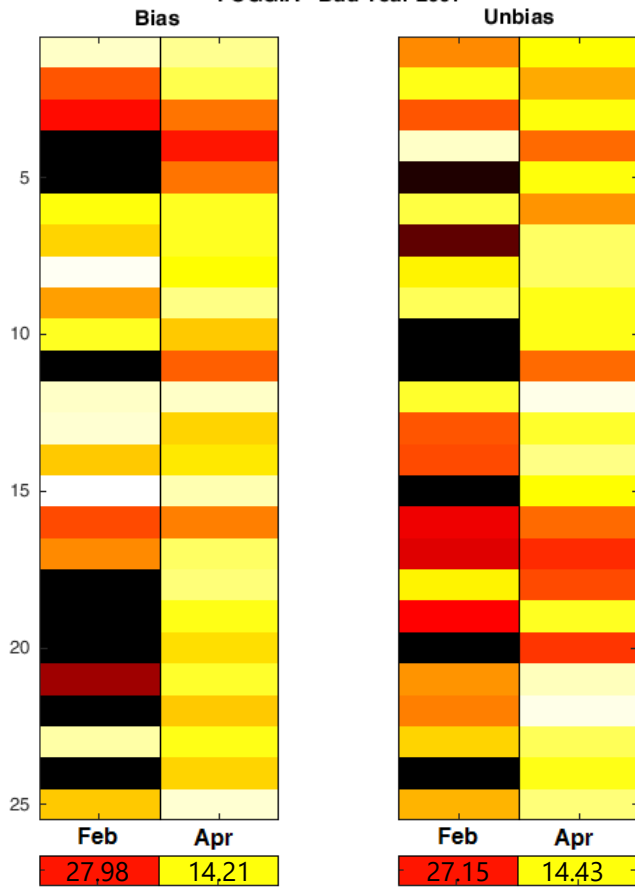


	DRY	AVG	WET
<b>Feb</b>	98.28	6.70	13.51
<b>Apr</b>	21.75	16.33	21.18

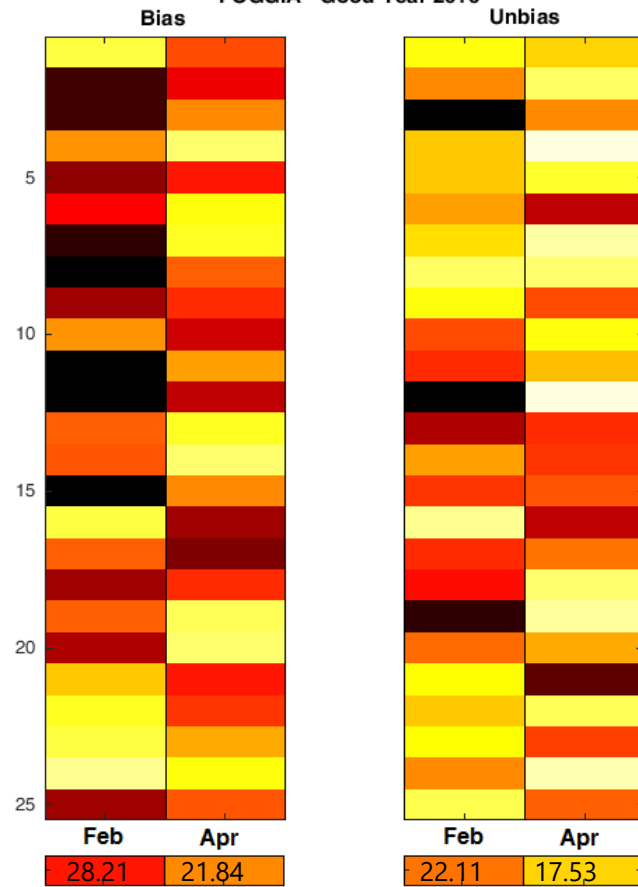
	DRY	AVG	WET
<b>Feb</b>	87.75	70.51	67.01
<b>Apr</b>	41.15	22.71	12.13



FOGGIA - Bad Year 2007

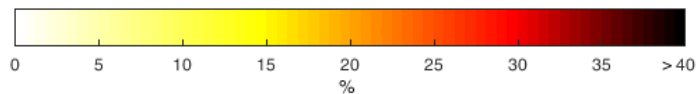


FOGGIA - Good Year 2016



	DRY	AVG	WET
<b>Feb</b>	75.96	4.36	7.15
<b>Apr</b>	6.10	6.90	5.53

	DRY	AVG	WET
<b>Feb</b>	71.16	31.25	25.30
<b>Apr</b>	11.83	10.69	11.32



# Final remarks and future challenges

- In general, the availability of unbiased data slightly improved the yield forecast, with the best result achieved for the high yielding crop year in Ancona, where 2 months before harvest the nRMSE dropped from 20.3% (*biased*) to 9.3% (*unbiased*).
- Based on these first promising results this benchmarking framework will be extended over a wider study area (the Italian Geographical Domain)
- A state-of-the-art skill model analysis will be performed to better highlight findings and remarks