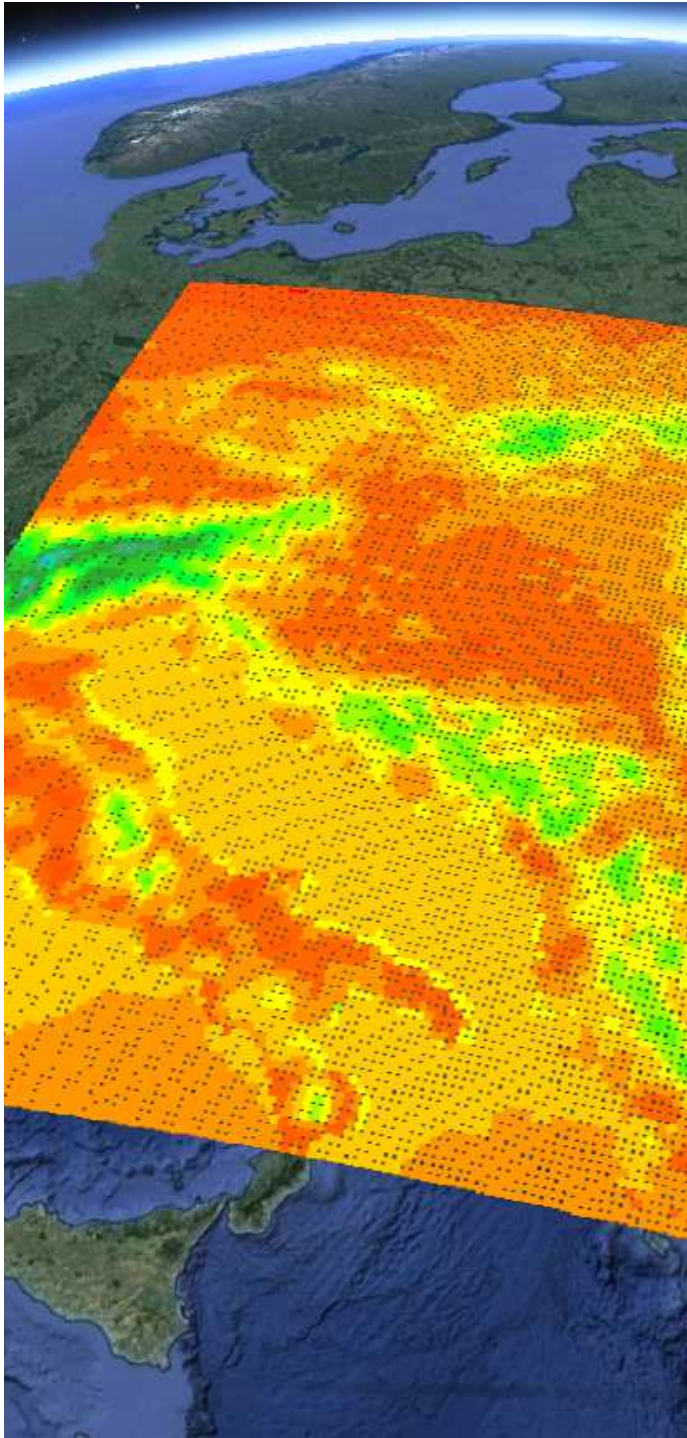




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H2020-TWINN-2015
AgMe⁺

Application of different NWP products in agricultural production: Limits and efficacy

Ana Firanj Sremac & Branislava Lalic
Faculty of Agriculture, UNS, Serbia



NWP products

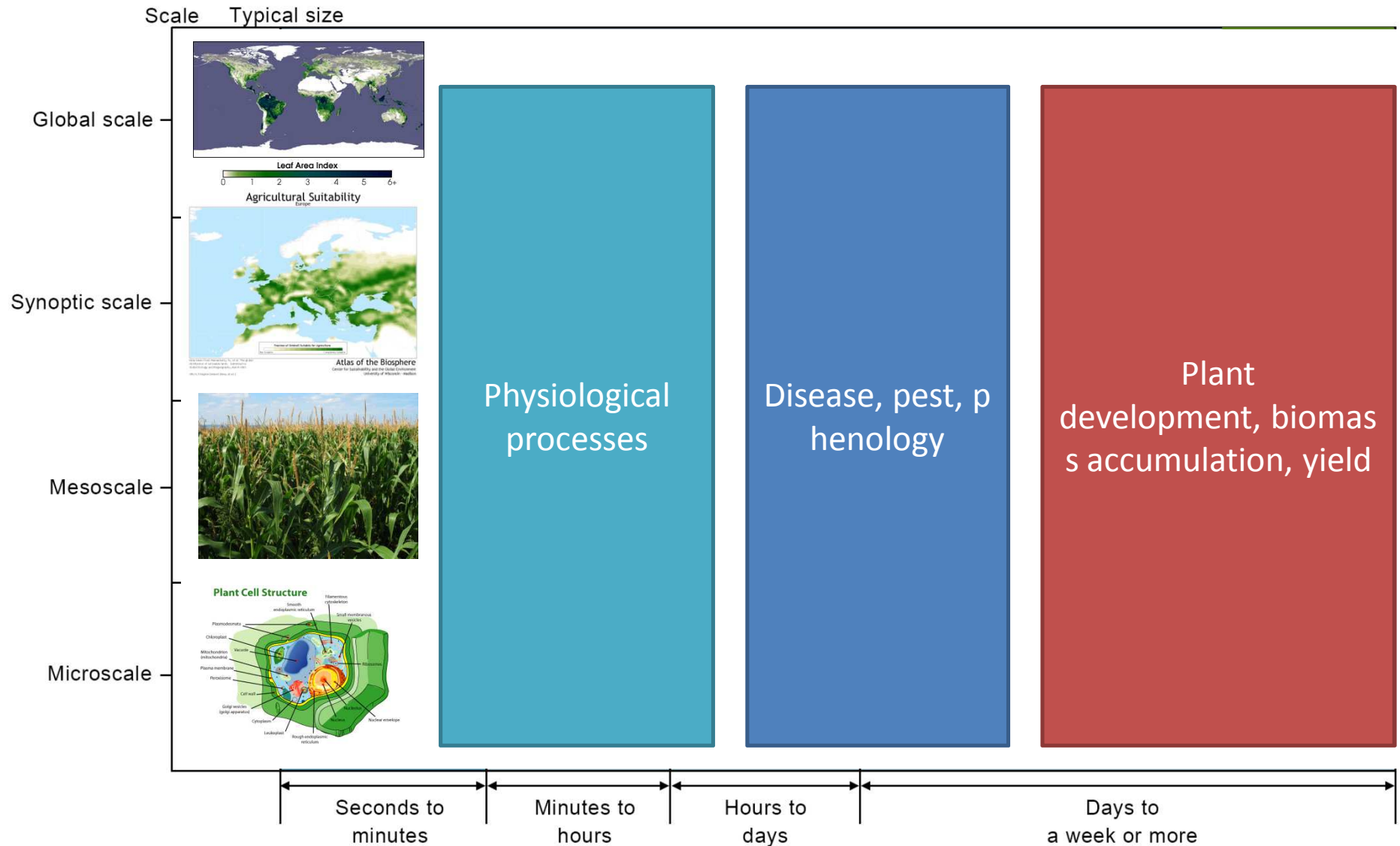
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1	Nowcasts	Descriptions of current weather parameters and descriptions of forecast weather parameters over the next 0-2-hours.
2	Very short-range	Descriptions of weather parameters over up to 12 hours into the future.
3	Short-range	Descriptions of weather parameters beyond 12 hours and up to 72 hours into the future.
4	Medium-range	Descriptions of weather parameters beyond 72 hours and up to 240 hours into the future.
5	Extended-range	Descriptions of weather parameters beyond 10 days and up to 30 days into the future. Typically averaged and expressed as departures from the climatological values for that period.
6	Long-range forecasts	From 30 days up to two years into the future.
6.1	Monthly outlooks	Descriptions of average weather parameters. Expressed as departures (deviations, variations, or anomalies) from the climatological values for that month (not necessarily the coming month).
6.2	Three-month or 90 day	Descriptions of the average values of weather parameters. Expressed as departures from the climatological values for that 90-day period (not necessarily the coming 90-day period).
6.3	Seasonal outlooks	Descriptions of the average values of weather parameters. Expressed as departures from the climatological values for that season.
7	Climate forecasts	Beyond two years.
7.1	Climate variability predictions	Descriptions of the expected values of climate parameters associated with the variations in inter-annual, decadal and multi-decadal climate anomalies.
7.2	Climate predictions	Descriptions of expected future climate conditions, including the effects of both natural and human driving factors.

Time and spatial scale

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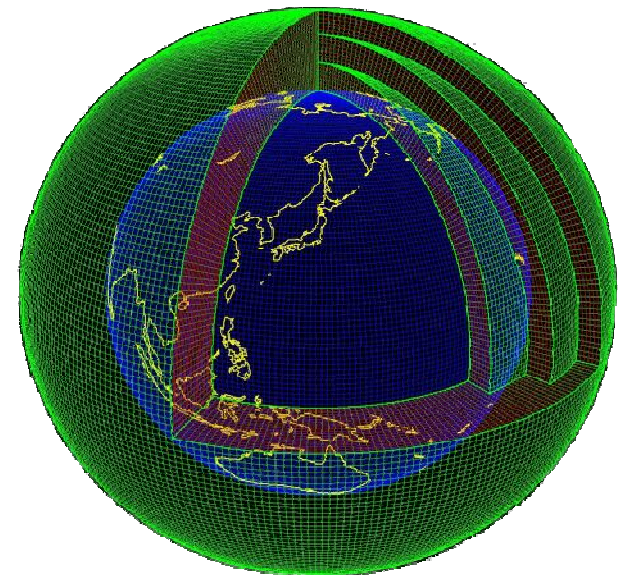


Used NWP products

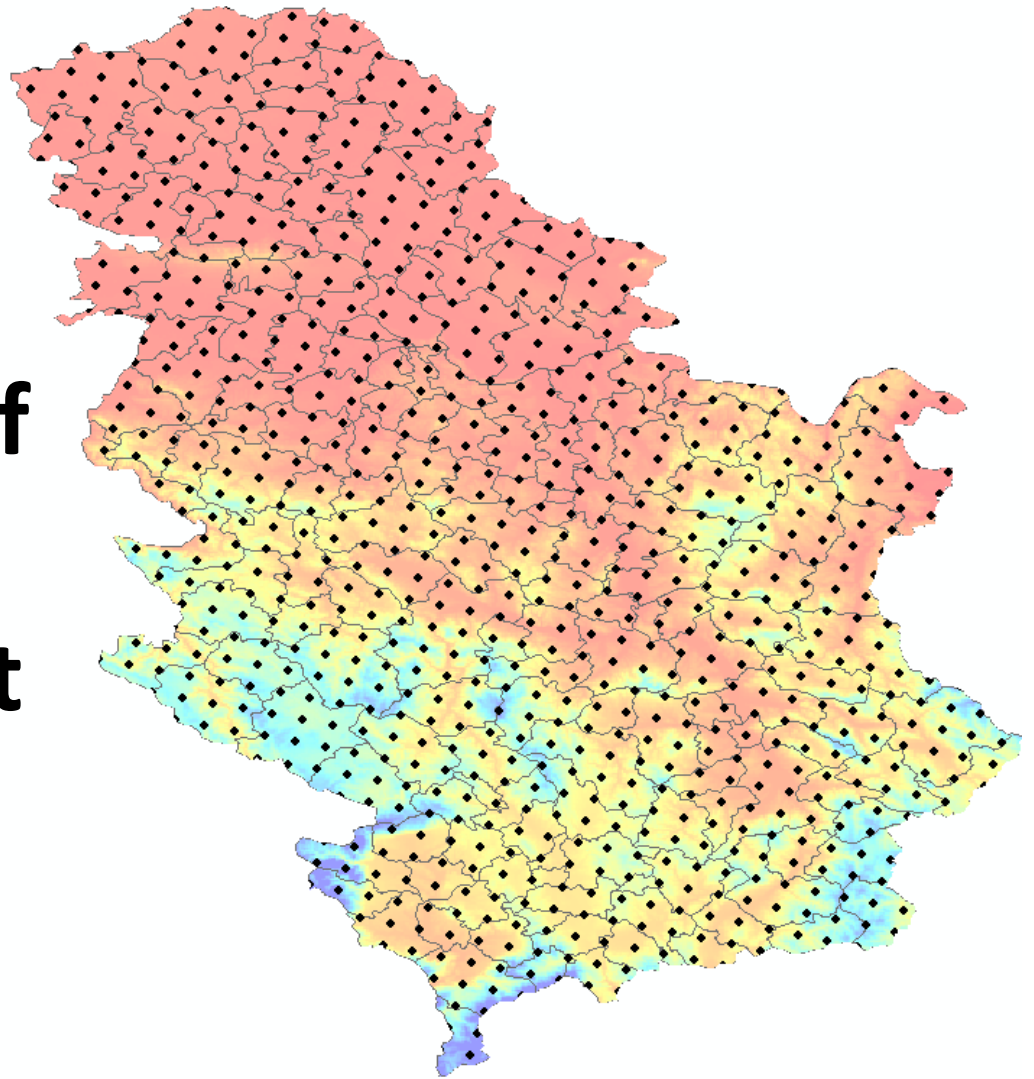
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- Short-range
 - Source NWP models: WorkEta, WRF-ARW
- Long-range forecasts (monthly, seasonal)
 - Source: ECMWF



An application of short-range weather forecast





- Select phenomena with adequate time step
 - plant diseases and pest appearance, phenology, frost
- Select adequate numerical tool
 - Disease and pest models, phenological models

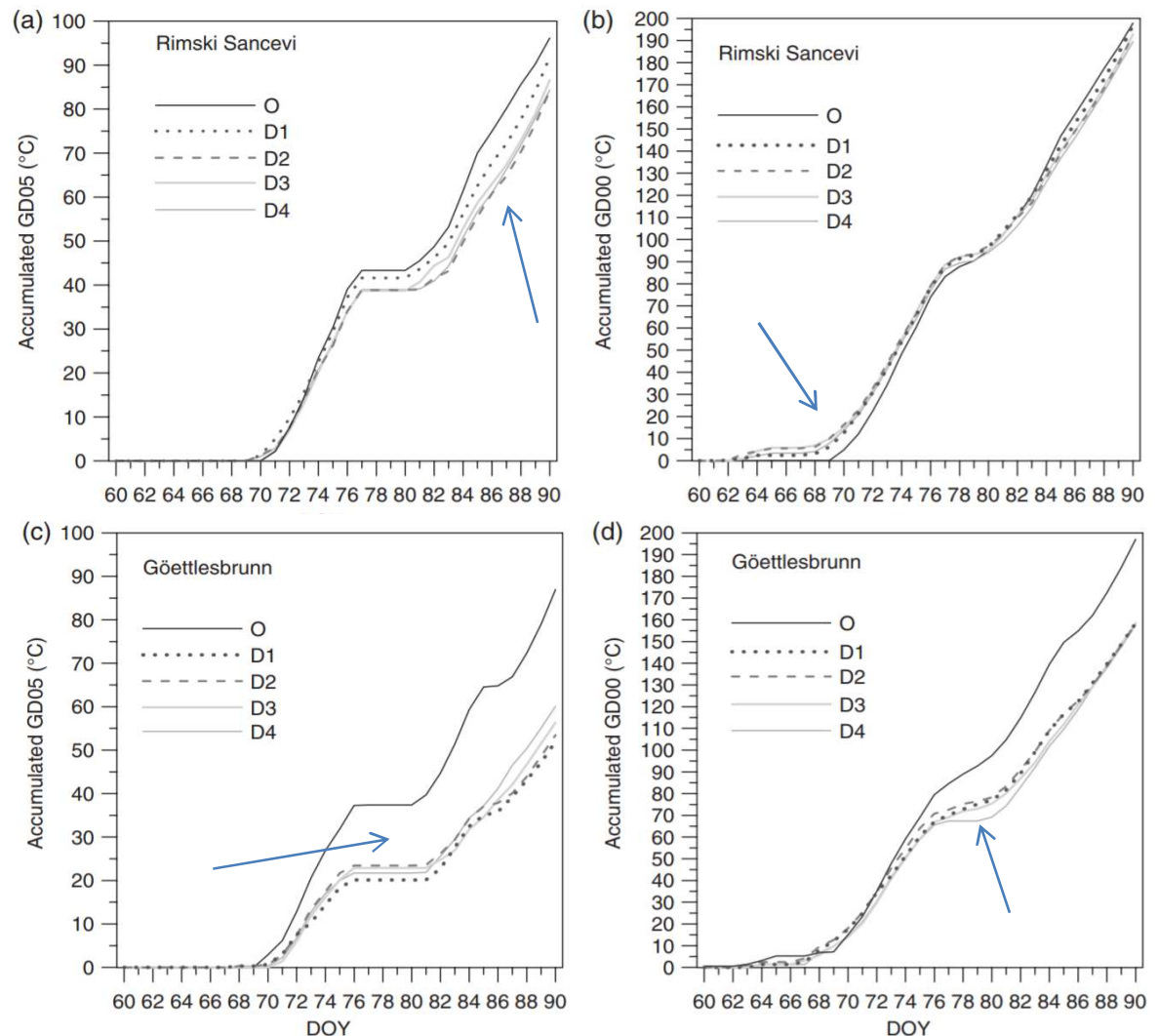
An application of short-range weather forecast

SERBIA FOR EXCELL



- **Lalic et al. (2016)**
Effectiveness of short-term numerical weather prediction in predicting **growing degree days** and meteorological conditions for **apple scab appearance**. *Meteorol. Appl.* 23: 50–56.
- NWP used: TheWorkEta model, 4 day, 10 km resolution
- Growing dynamics, Disease model: BAHUS
 - **Limit** – computational time
 - **Efficiency** – small difference between 1,2,3,4 day forecast

Accumulated growing degree days

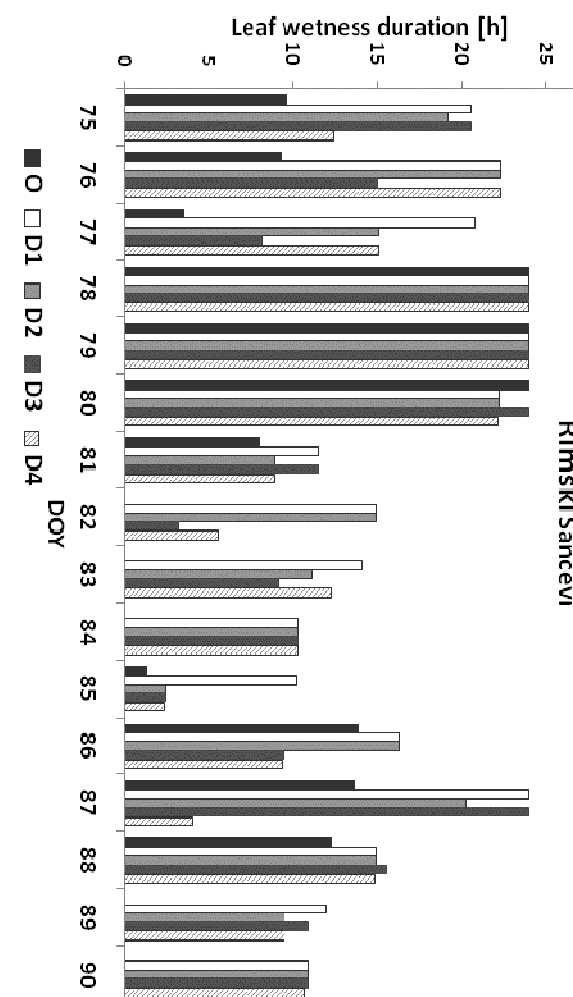
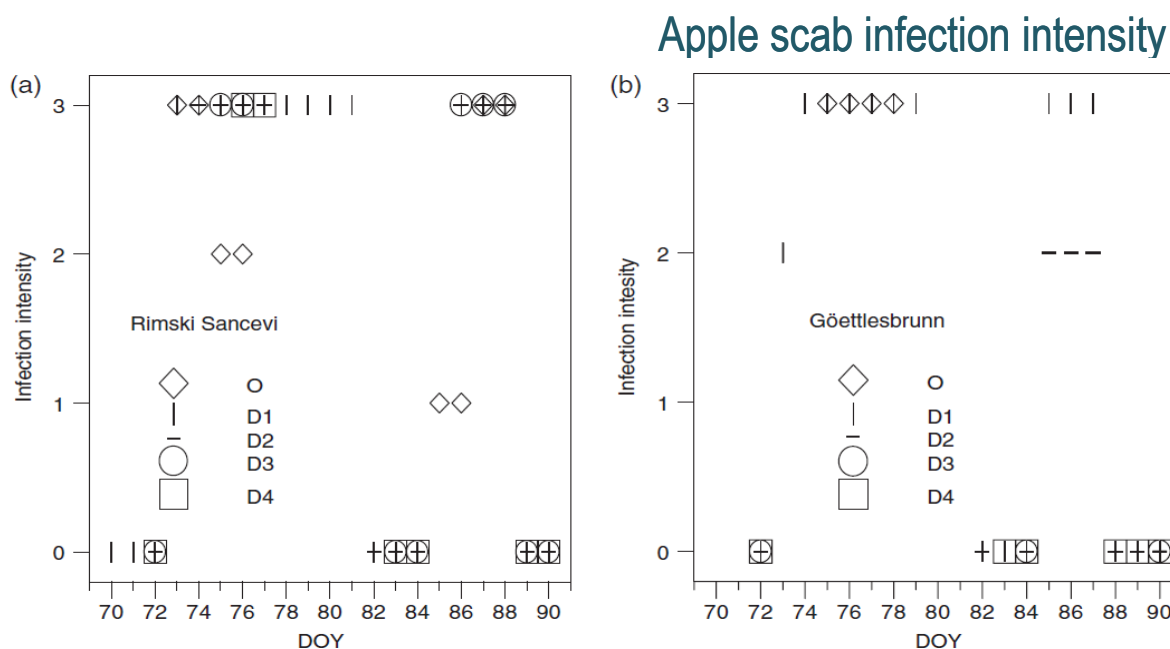


An application of short-range weather forecast

SERBIA FOR EXCELL



- **Lalic et al. (2016)** Effectiveness of short-term numerical weather prediction in predicting **growing degree days** and meteorological conditions for **apple scab appearance**. Meteorol. Appl. 23: 50–56.
- NWP used: TheWorkEta model, 4 day, 10 km resolution
- Growing dynamics, Disease model: BAHUS
Limit – computational time, orography
Efficiency – small difference between 1,2,3,4 day forecast

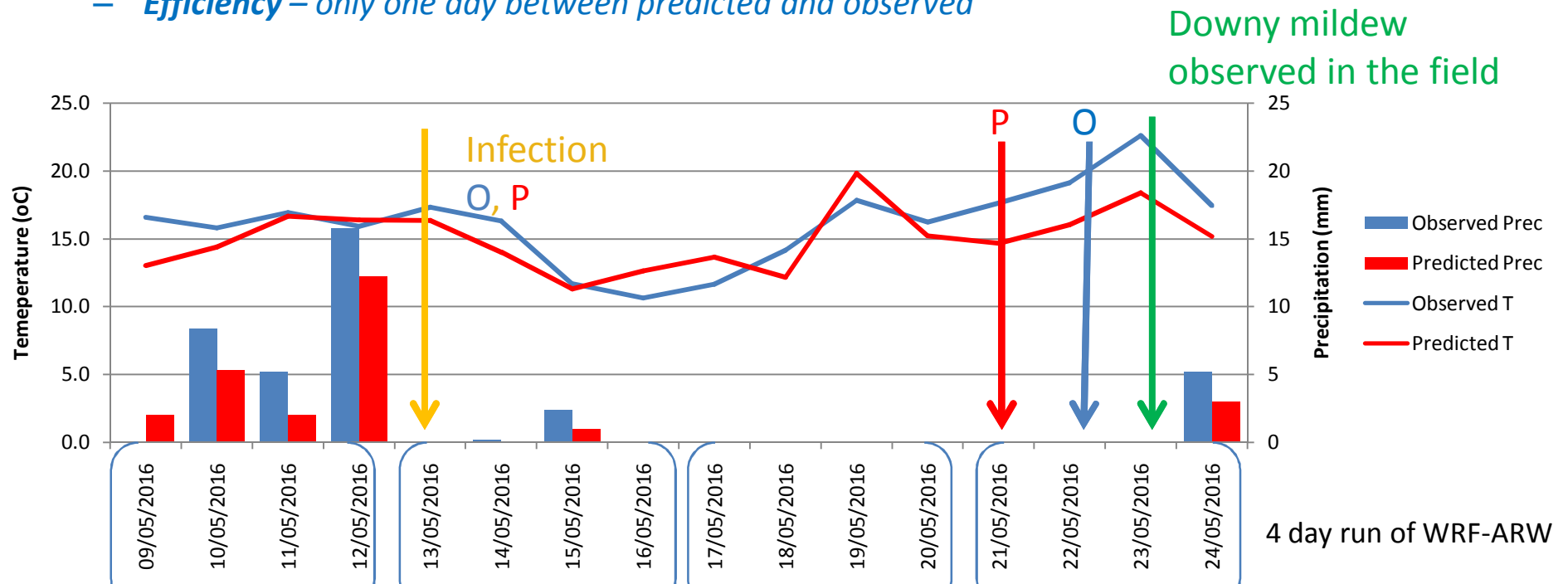


An application of short-range weather forecast

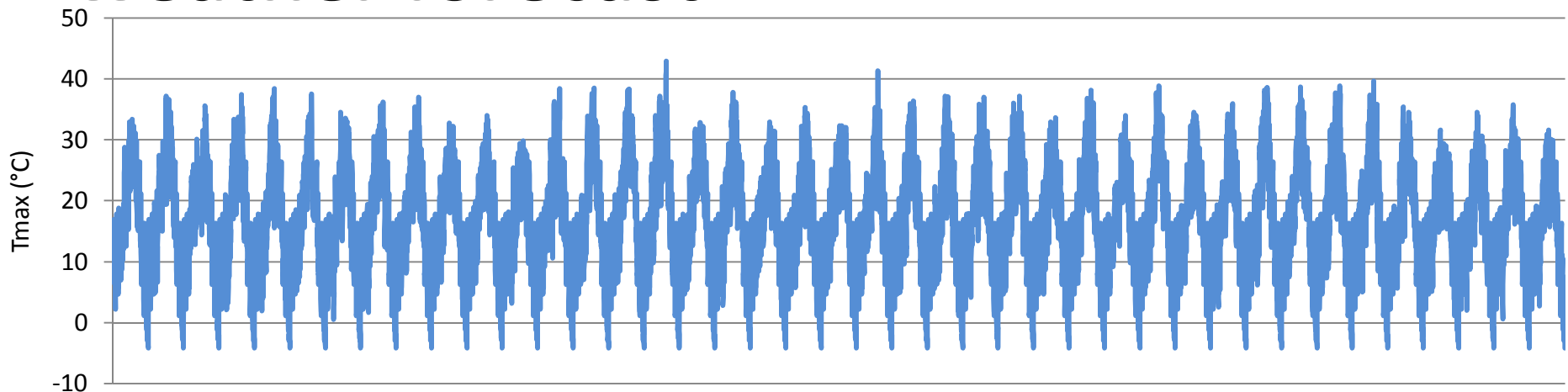
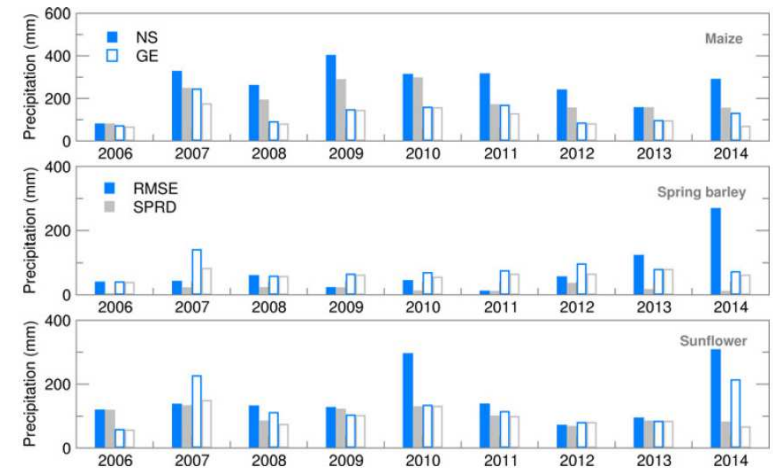
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- **Firanj Sremac et al. (2016)** The WRF-ARW application in predicting meteorological conditions for **Downy mildew** (*Plasmopara viticola*) appearance of wine grape. Abstract EMS, Trieste, Italy, 12-16 September 2016
- NWP used: WRF-ARW model, 4 day, 10 km
- Disease model: BAHUS (**Müller's method**)
 - *Limit* – computational time, errors in precipitation
 - *Efficiency* – only one day between predicted and observed



An application of monthly and seasonal weather forecast

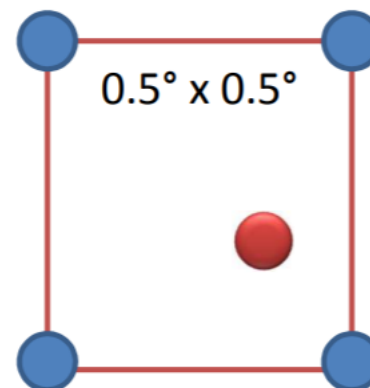
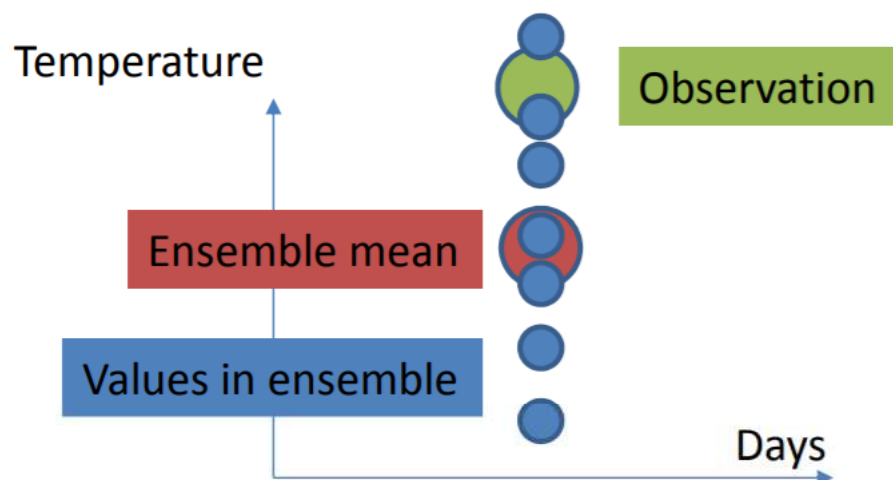




- Select phenomena with adequate time step
 - Plant phenology (monthly), vegetation dynamics, biomass, yield
- Select adequate numerical tool
 - Crop model
- Ensemble forecast -> ensemble of phenology/
crop model outputs



Data sets – ensemble forecast



Observed average temperature for March 2017: 10.5 °C

Forecasted – control run temperature for March 2017: 10.8 °C

Forecasted – ensemble for March 2017, averaged by ensemble:

1. 10.1	5. 10.	9. 10.2
2. 11	6. 12.1	10. 11.1
3. 10.4	7. 10.5	11. 10.5
4. 10.9	8. 10.6	12. 10.1

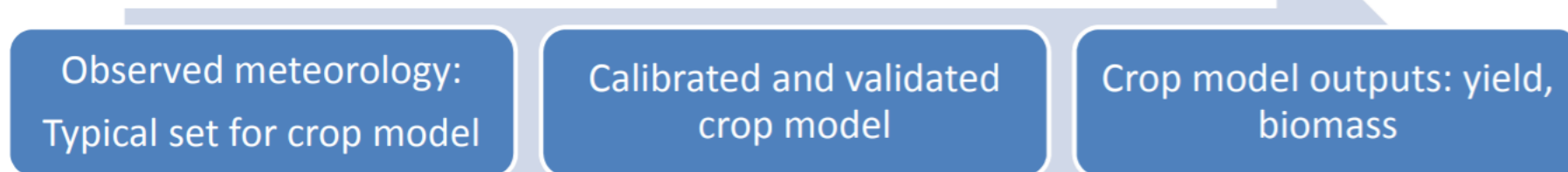
Ensemble mean = 10.6 °C

An application of monthly and seasonal weather forecasting

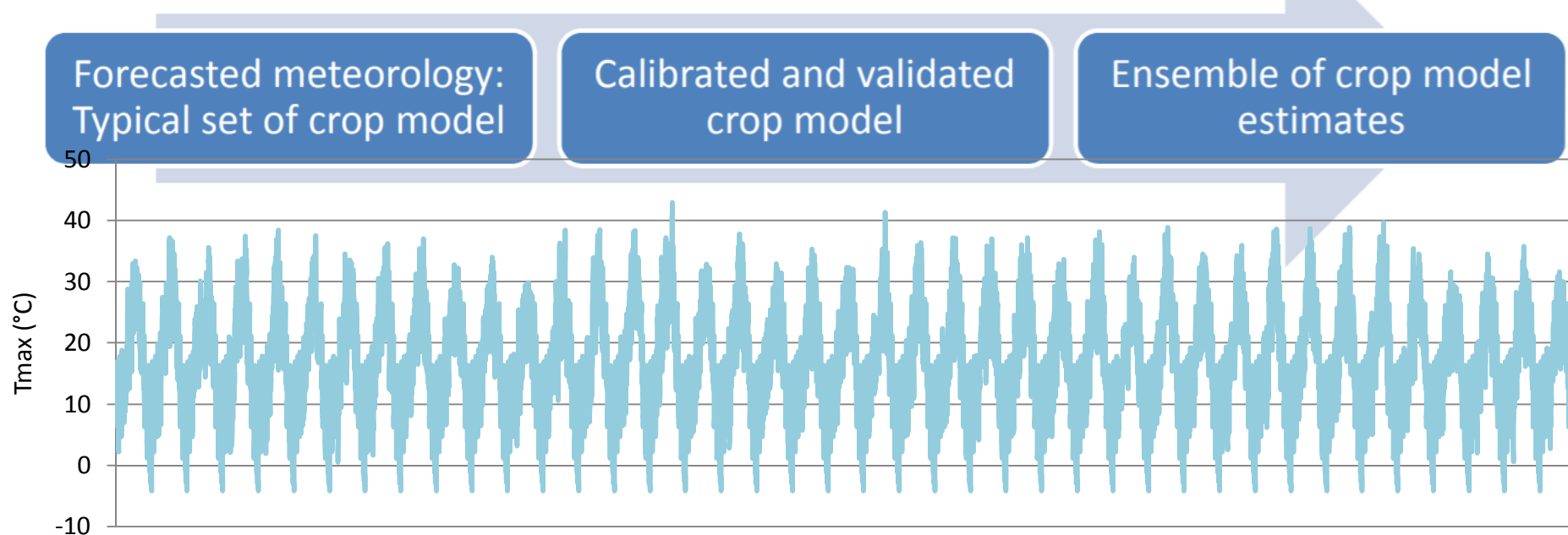
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First run – observed meteorology



One run per one ensemble member





VERIFICATION STATISTICS

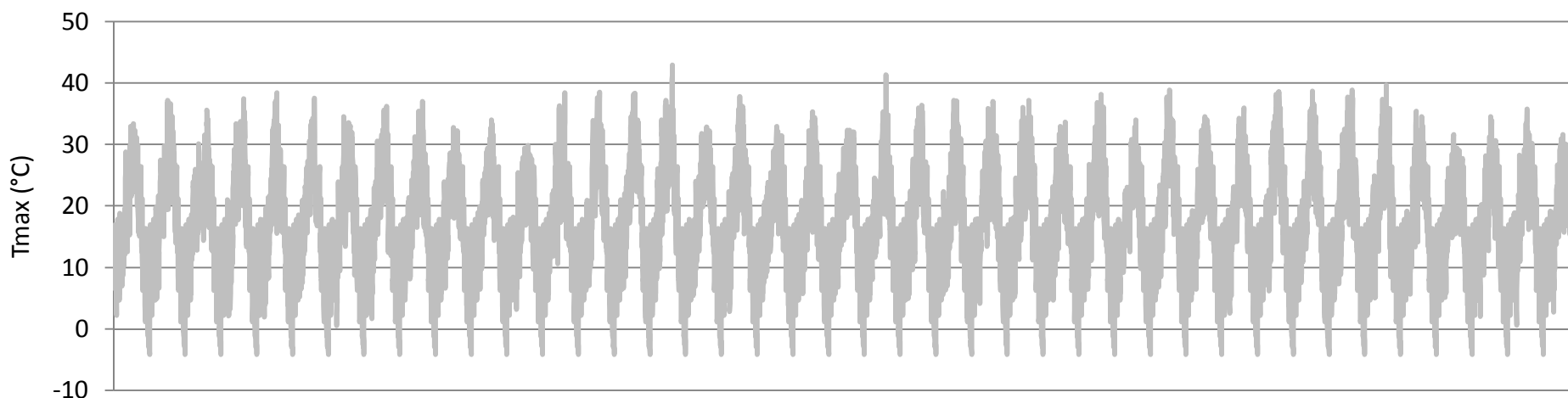
$$RMSE = \frac{1}{N} \sqrt{\sum_{i=1}^N (\bar{A} - A_{OBS})^2}$$

$$SPREAD = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (\bar{A} - A(i))^2}$$

Ensemble statistics

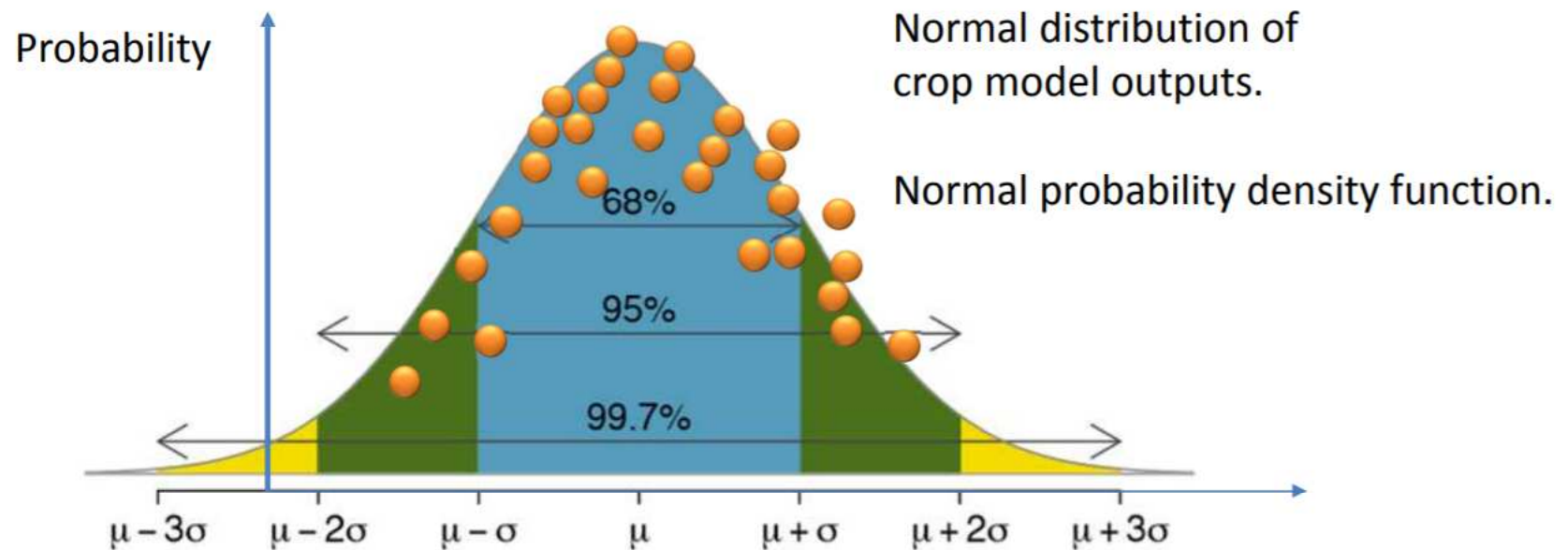
$$RLD = \frac{|A_{CR} - A_{OBS}|}{A_{OBS}} \cdot 100\%$$

Control run statistics





IGNORANCE SCORE



$$I(p(y), Y) = -\log_2(p(Y))$$

$p(Y)$ is a unitless probability density function of CMO

After calculating Ignorance for different crop model outputs, we can compare them.

Ignorance is < 2.04 , the model is very good, and if it is > 7.81 , the model is not adequate.

An application of monthly weather forecasting

SERBIA FOR EXCELL



- Lalic et al. (2017a) Testing **efficacy of monthly forecast** application in agrometeorology: Winter wheat **phenology dynamic**. *IOP Conf. Ser.: Earth Environ. Sci.* 57 012002
- ECMWF – monthly forecast, MARCH-JUNE, 51 ensemble, 35 km resolution
- Crop model SIRIUS
 - **Limit** – dependant on the position of the grid points
 - **Efficiency** – small difference RMSE and SPRED for Tmin, Tmax, Yield, RS

Setting	RMSE (°C)				SPREAD (°C)			
	M	A	M	J	M	A	M	J
T_{min}								
Groß-Enzersdorf	2.9	1.6	1.0	0.8	1.2	1.1	0.9	0.7
Rimski Sancevi	1.3	1.2	1.0	1.6	1.2	1.2	0.9	0.7
T_{max}								
Groß-Enzersdorf	5.1	4.0	2.4	2.8	1.5	1.4	1.3	1.3
Rimski Sancevi	3.0	1.9	1.7	1.5	1.7	1.4	1.4	1.5

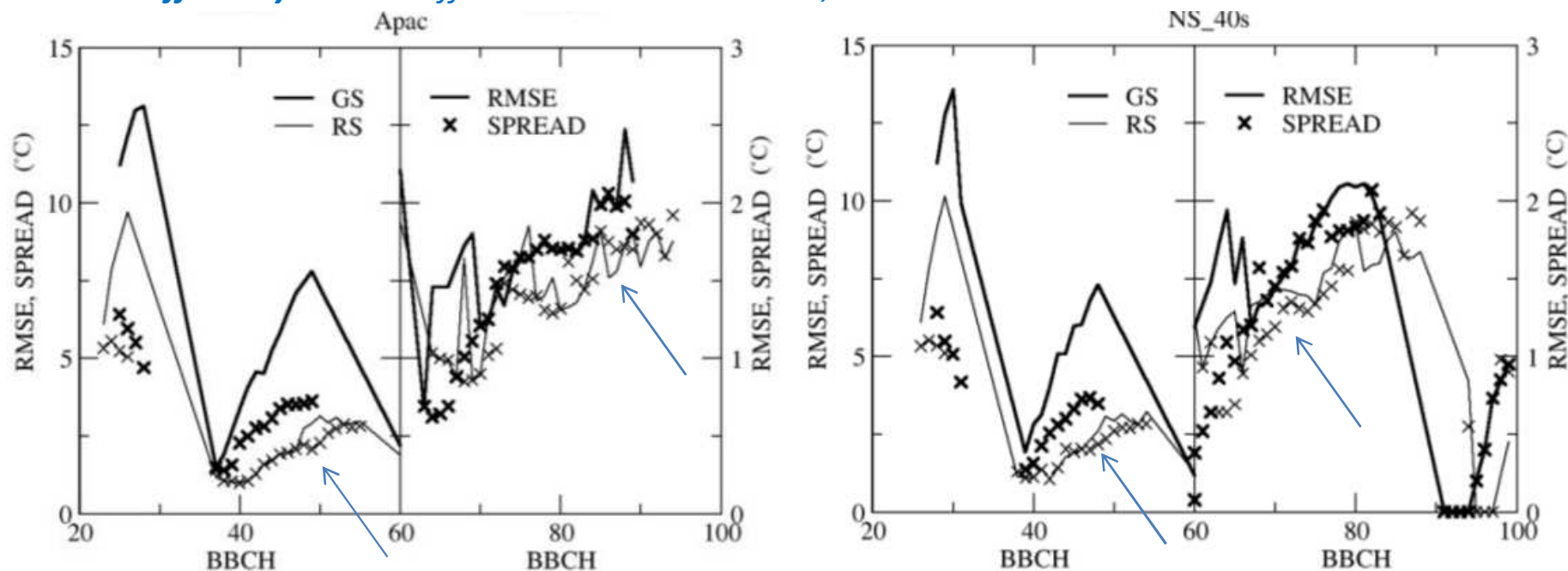
	ET (mm)	Max def. (mm)	Anthesis day	Maturity day	Biomass (t/ha)	Yield (t/ha)
Grossensdorf obs	395	187	149	190	14.859	6.989
Grossensdorf month	355	145	149	187	13.899	5.940
Novi Sad obs	425	136	139	181	14.473	6.006
Novi Sad month	379	143	139	179	14.478	5.920

An application of monthly weather forecasting

SERBIA FOR EXCELL

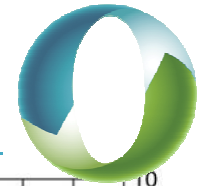


- Lalic et al. (2017a) Testing **efficacy of monthly forecast** application in agrometeorology: Winter wheat **phenology dynamic**. *IOP Conf. Ser.: Earth Environ. Sci.* 57 012002
- ECMWF – monthly forecast, MARCH-JUNE, 51 ensemble, 36 km resolution
- Crop model SIRIUS
 - **Limit** – dependant on the position of the grid points
 - **Efficiency** – small difference RMSE and SPRED, RS

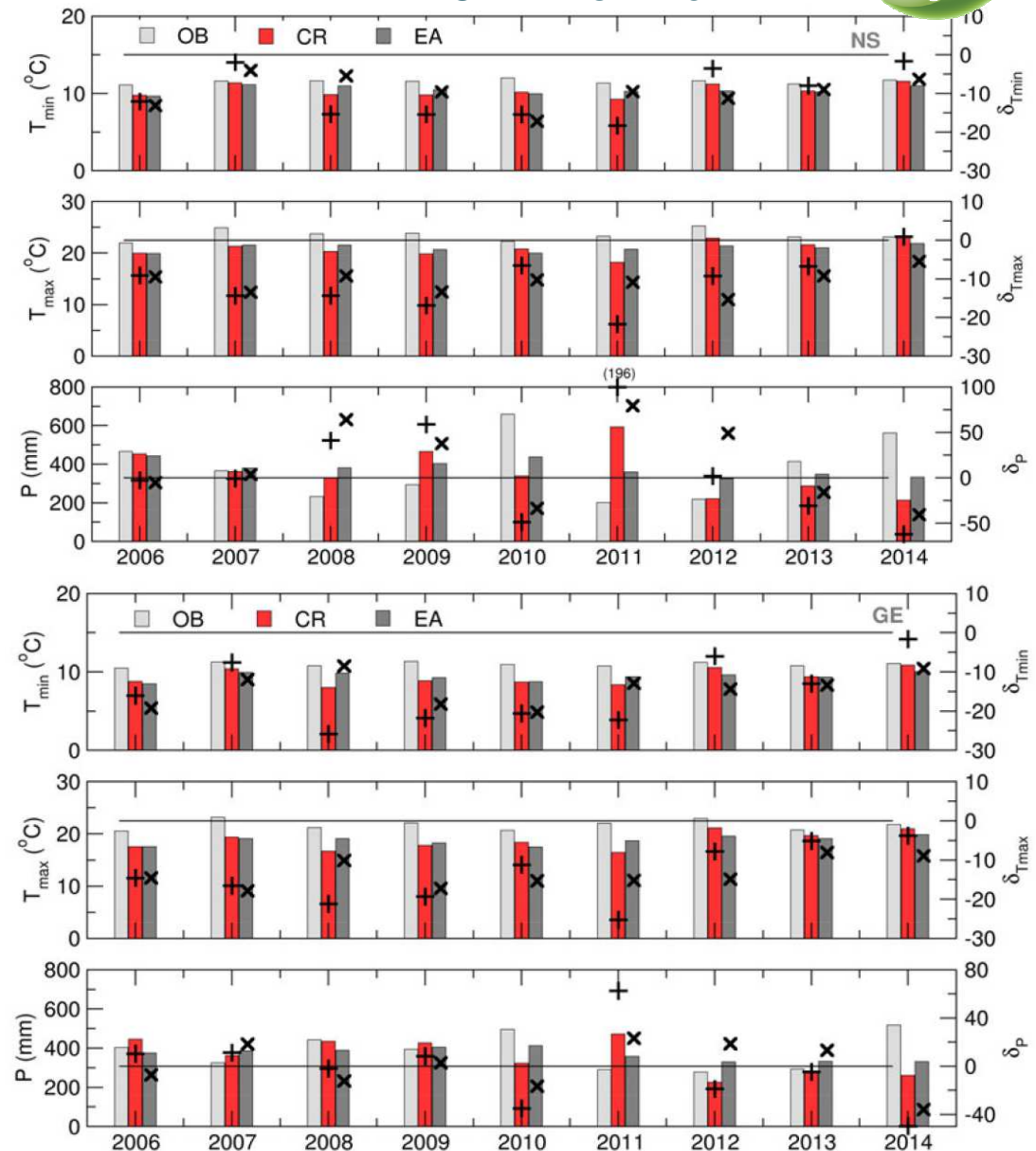


An application of seasonal weather forecasting

SERBIA FOR EXCELL



- Lalic et al. (2017b) Seasonal forecasting of **green water components** and **crop yields of winter wheat** in Serbia and Austria. *Journal of Agricultural Science*, 1-17
- ECMWF – seasonal forecast, MARCH-OCTOBER, 10-50 ensemble (depending on the year), $0.5^\circ \times 0.5^\circ$ resolution
- Crop model: SIRIUS
 - **Limit** – dependant on the position of the grid points, OB is larger than EA, CR
 - **Efficiency** – small difference RMSE and SPRED

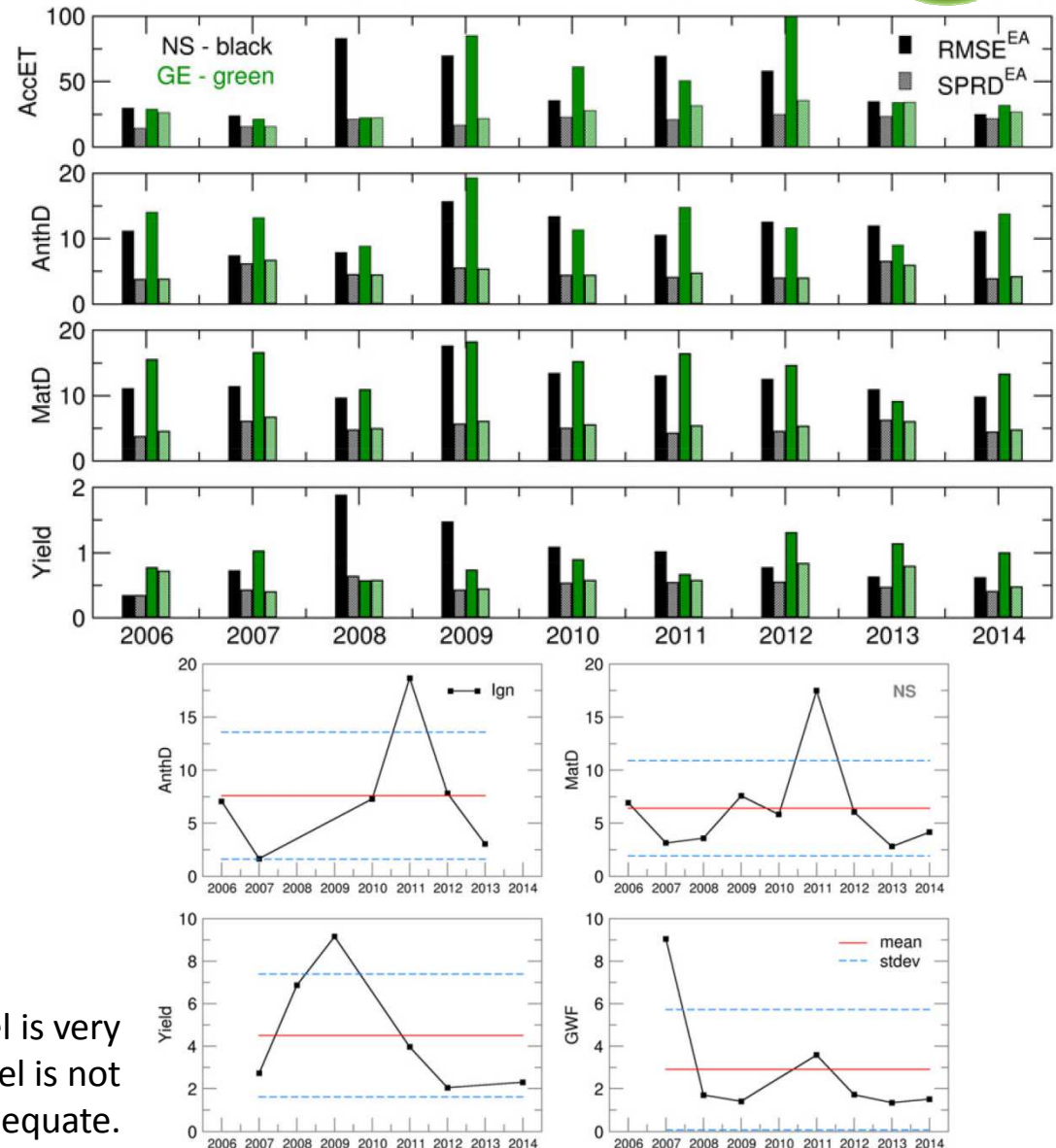


An application of seasonal weather forecasting

SERBIA FOR EXCELL



- Lalic et al. (2017b) Seasonal forecasting of **green water components** and **crop yields of winter wheat** in Serbia and Austria. *J. Agric.Sci*, 1-17
 - ECMWF – seasonal forecast, MARCH-OCTOBER, 10-50 ensemble (depending on the year), $0.5^\circ \times 0.5^\circ$ resolution
 - Crop model: SIRIUS
 - **Limit** – dependant on the position of the grid points
 - **Efficiency** – small difference RMSE and SPRD for Yield
- Ignorance is < 2.04 , the model is very good, and if it is > 7.81 , the model is not adequate.



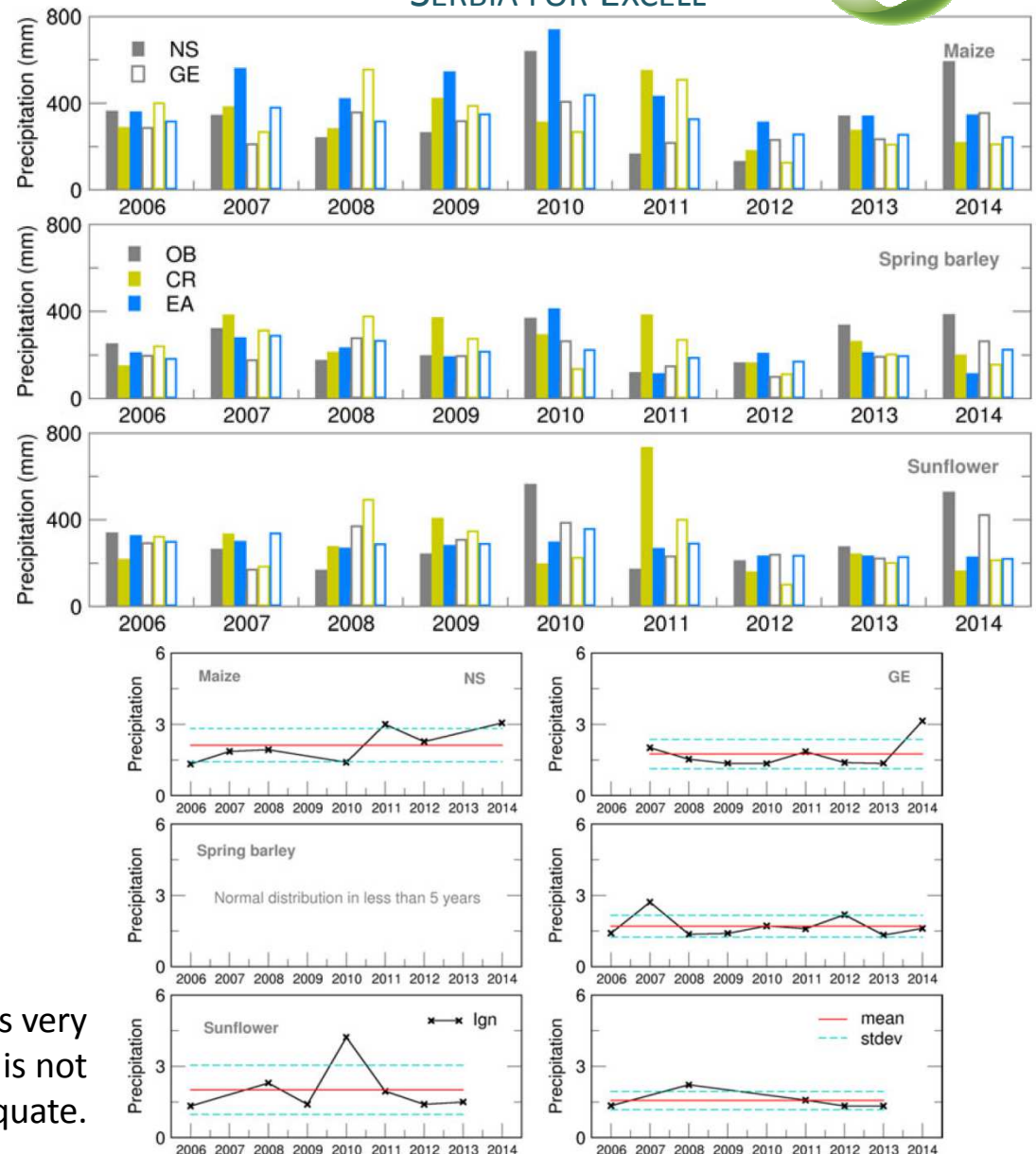
An application of seasonal weather forecasting



SERBIA FOR EXCELL

- Lalic et al. (2018) Seasonal forecasting of **green water components** and crop yield of summer crops in Serbia and Austria. *J. Agric.Sci*, 1-15
- ECMWF – seasonal forecast, MARCH-OCTOBER, 10-50 ensemble (depending on the year), $0.5^\circ \times 0.5^\circ$ resolution
- Crop model: AquaCrop
 - **Limit** – dependant on the position of the grid points, but in years with extremes EA closer to OB
 - **Efficiency** – average Ignorance less then 3.

Ignorance is < 2.04, the model is very good, and if it is > 7.81, the model is not adequate.



An application of seasonal weather forecasting

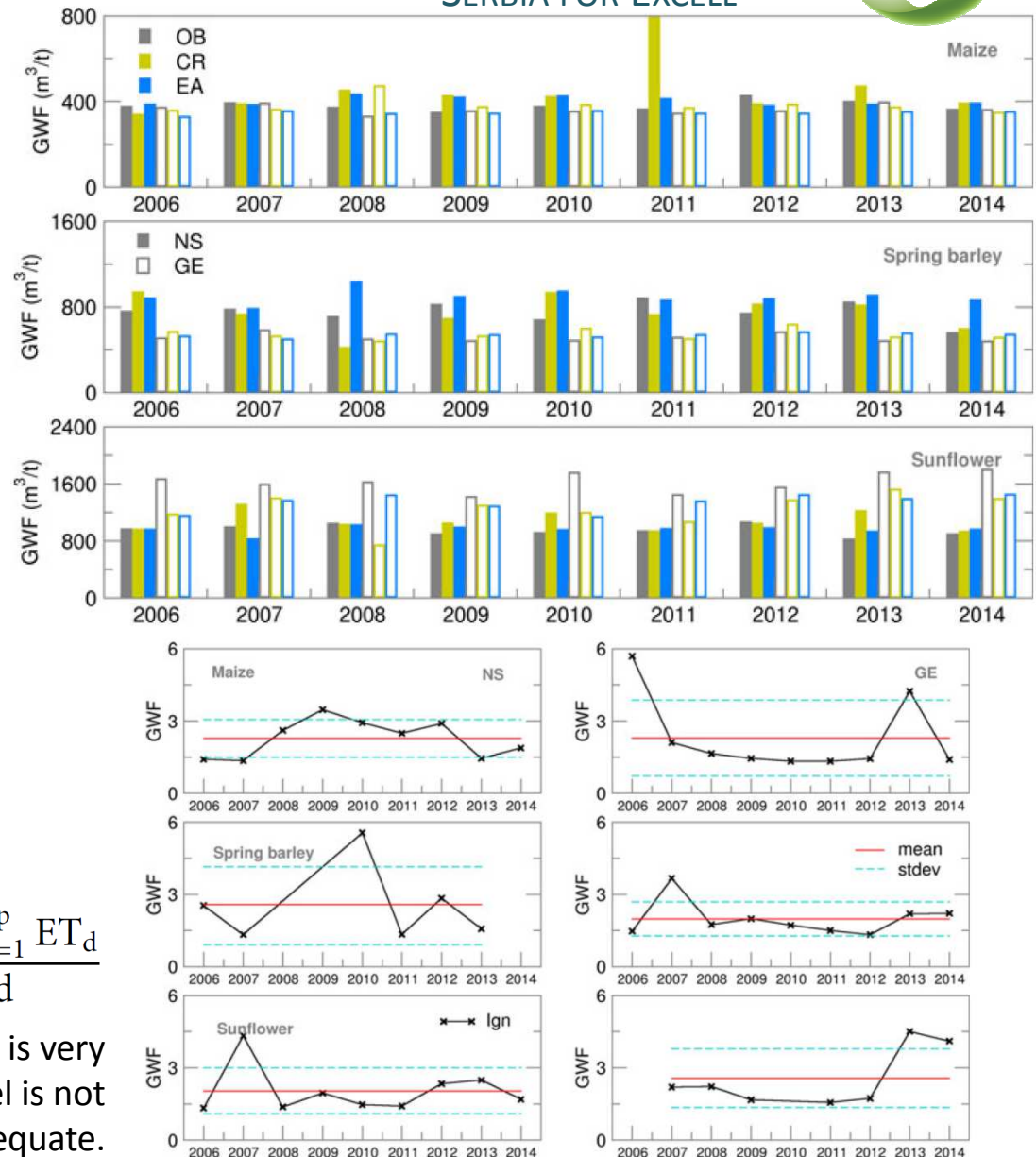


SERBIA FOR EXCELL

- Lalic et al. (2018) Seasonal forecasting of **green water components** and **crop yield** of summer crops in Serbia and Austria. *J. Agric.Sci*, 1-15
- ECMWF – seasonal forecast, MARCH-OCTOBER, 10-50 ensemble (depending on the year), 0.5° × 0.5° resolution
- Crop model: AquaCrop
 - **Limit** – dependant on the position of the grid points, and cultivar
 - **Efficiency** – apart from some years in average Ign less than 3

$$GWF = \frac{10 \times \sum_{d=1}^{l_{gp}} ET_d}{Yield}$$

Ignorance is < 2.04, the model is very good, and if it is > 7.81, the model is not adequate.



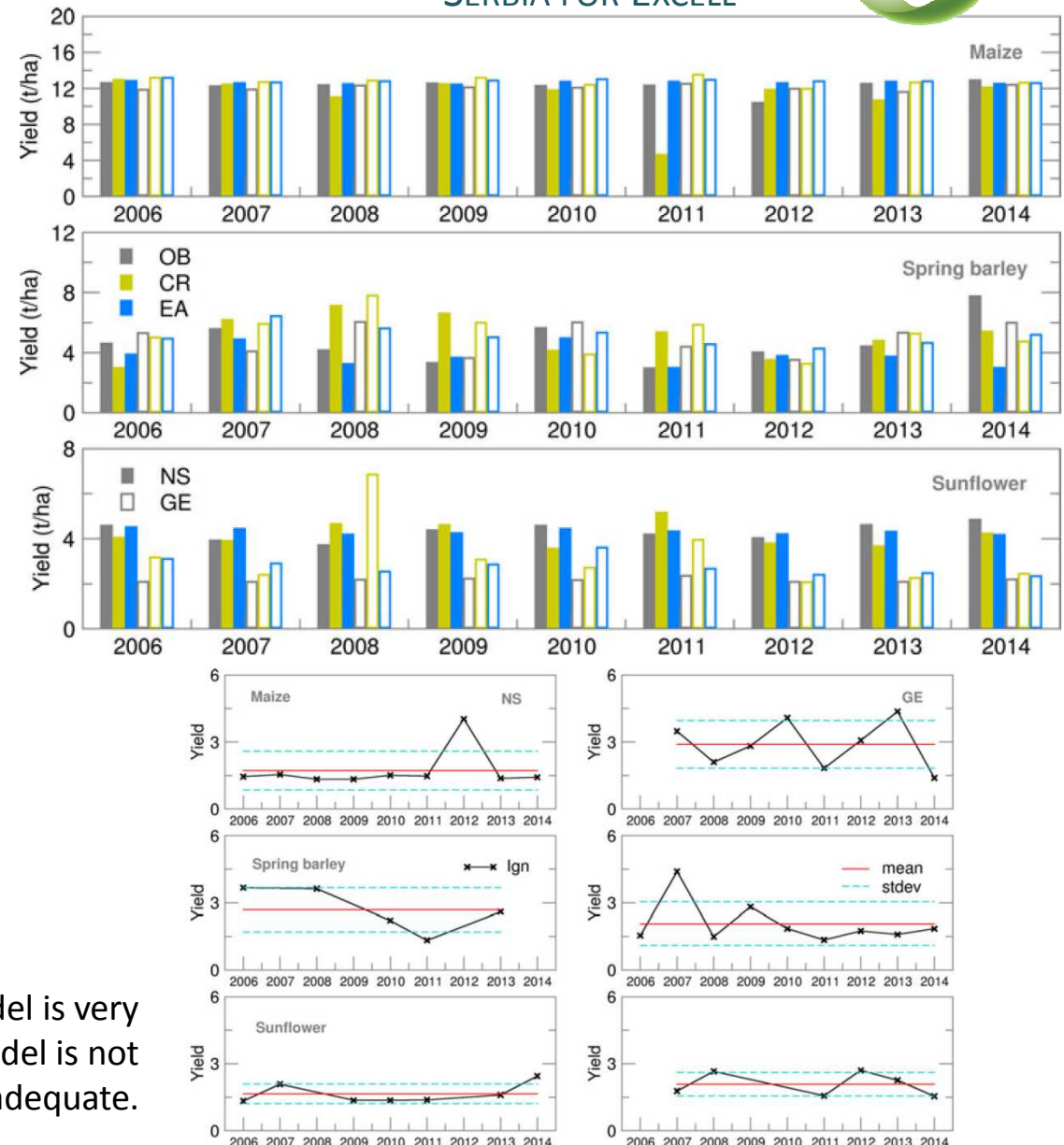
An application of seasonal weather forecasting



SERBIA FOR EXCELL

- Lalic et al. (2018) Seasonal forecasting of green water components and **crop yield** of summer crops in Serbia and Austria. *J. Agric.Sci*, 1-15
- ECMWF – seasonal forecast, MARCH-OCTOBER, 10-50 ensemble (depending on the year), $0.5^\circ \times 0.5^\circ$ resolution
- Crop model: AquaCrop
 - **Limit** – dependant on the position of the grid points, and cultivar
 - **Efficiency** – average Ignorance less than 3 for all

Ignorance is < 2.04, the model is very good, and if it is > 7.81, the model is not adequate.



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Thank you