18 November 2020

Barcelona Supercomputing Center
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EXCELENCIA SEVERO OCHOA

Forecast quality assessment for operational climate prediction

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with contributions from Joaquín Bedia, Nicola Cortesi, Nube González, Carlo Lacagnina, Llorenç Lledó, Raül Marcos, Núria Pérez, Jaume Ramón, Albert Soret, Marta Terrado, Verónica Torralba, Ilaria Vigo and many more

S2S4E Climate Services for Clean Energy

EUCP European Climate Prediction system

ICREA

Climate information requirements

- Salience: *It refers to the relevance of information for an actor's decision choices*. Often interesting scientific questions are far from a real-world situation.

- Credibility: *It refers to whether an actor perceives information as meeting standards of scientific plausibility and technical adequacy*. Sources must be trustworthy and/or believable.

- Legitimacy: *It refers to whether an actor perceives the climate information process as unbiased and meeting standards of political and procedural fairness*.

Power

Reputation

Values

Transparency

Standards

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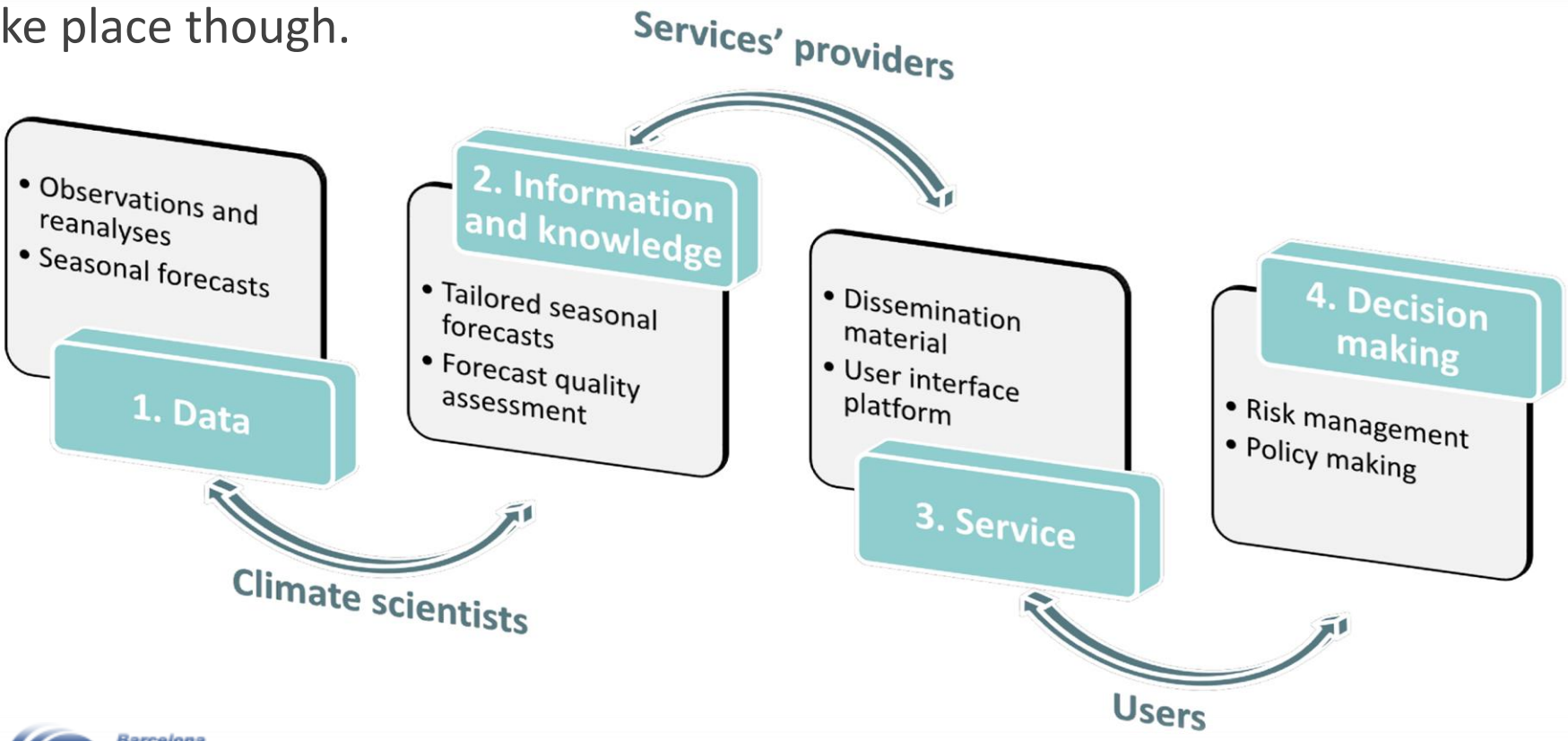
Cash et al. (2003, PNAS)

The research-provider-service paradigm

A service-oriented research agenda requires the traditional chain "research development-operations-service provision" to move both ways so that not only information quality is demonstrated, but user requirements are adequately addressed and value illustrated. This leaves a space for transdisciplinary research. This chain should not preclude basic research to take place though.



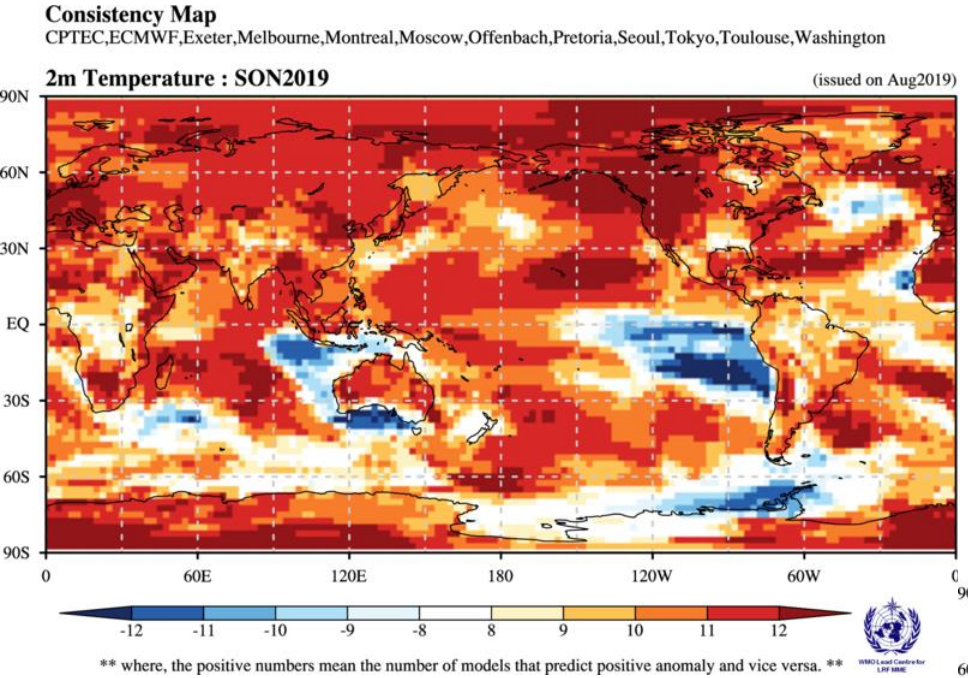
Forecast requirements identified by users

- Targeted products need to become widely available, easy to access and understood by different professionals
- Need to understand how the information provided can be used and integrated in operations and activities.
- Added value of predictions needs to be better understood. The chain of predictability source needs to be established.
- Need to reduce uncertainty. The skill is too low to base decisions on them, since the uncertainty is high.
- Better explanation of the link between operations and climate change projections.
- The information needs to be reliable enough.
- Maybe need for fine spatial resolutions or allow for urban areas.

Forecast quality is an integral part of the service to the user because it is inseparable from the forecast product

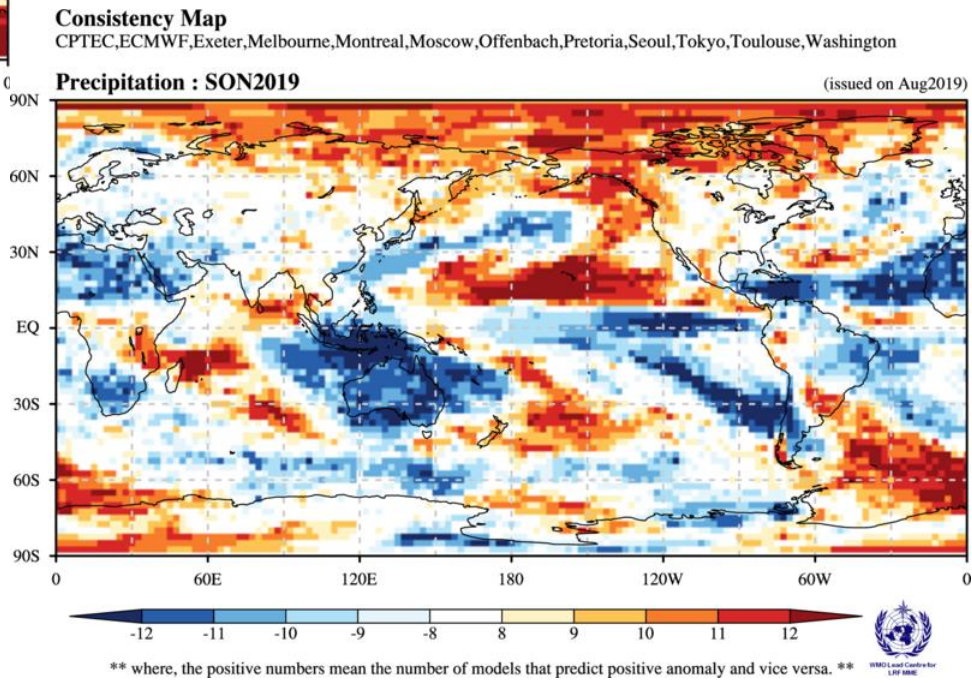
Extracted from project deliverables of EUCP (D6.4), PRIMAVERA (D11.6) and EUPORIAS (D12.3). Additional sectoral comments in user engagement by S2S4E, APPLICATE, MED-GOLD, HIATUS and VISCA.

Users look for climate forecasts online



But some elements are missing:

- Quality assurance
- Traceability
- Interpretation
- Authority



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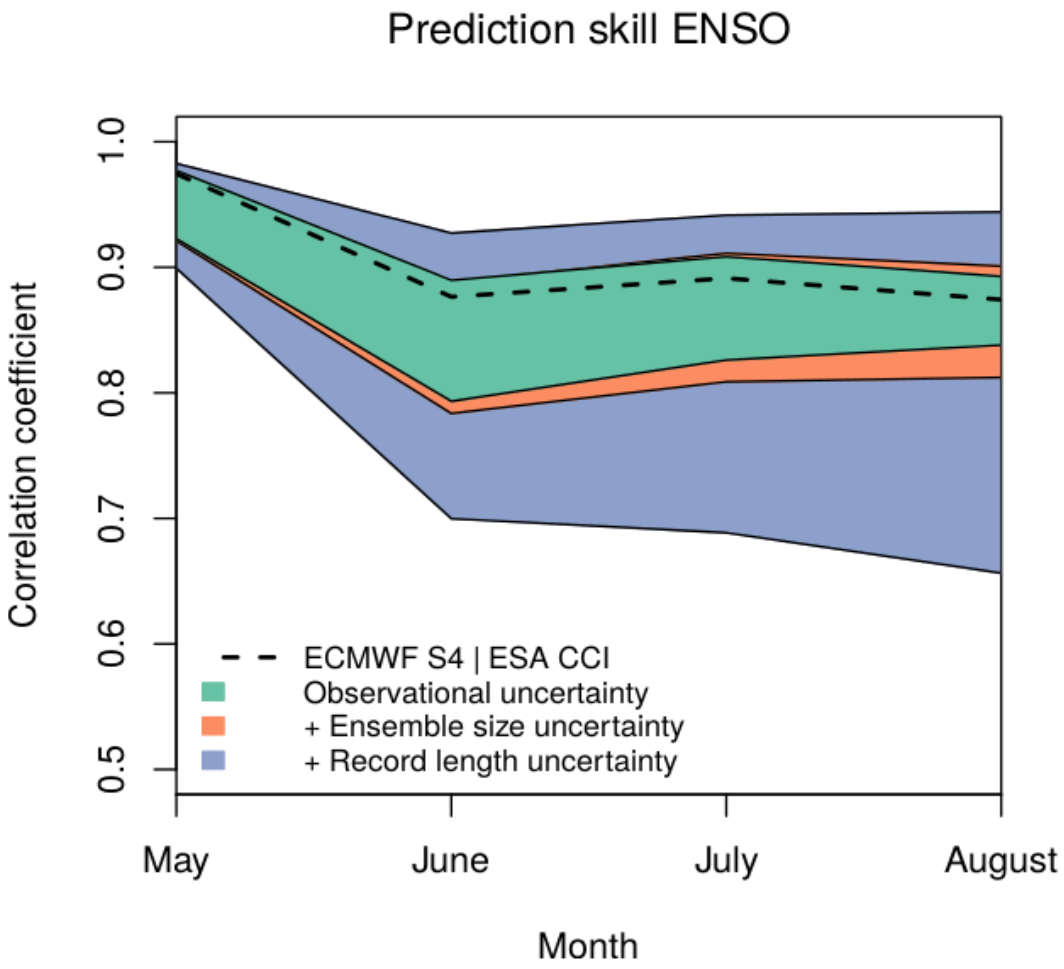
Some key elements for users

- Observational uncertainty: comparison between reanalyses in a forecast verification context.
- Definition of standard procedures: standards are less common than one would expect.
- Traceability and quality control: quality control and reproducibility of data and products is increasingly important in the research community, but its operational aspects are not solved yet.
- User indicators: indicators often do not have the same level of skill as the meteorological variables.
- Interpretation and communication: users are often not experts, and even when they are it is easy to misunderstand the existing information. Communication is a challenge
- Synthesis and narratives: how to deal with multiple lines of evidence in the message constructions.

Sources of uncertainty of forecast quality

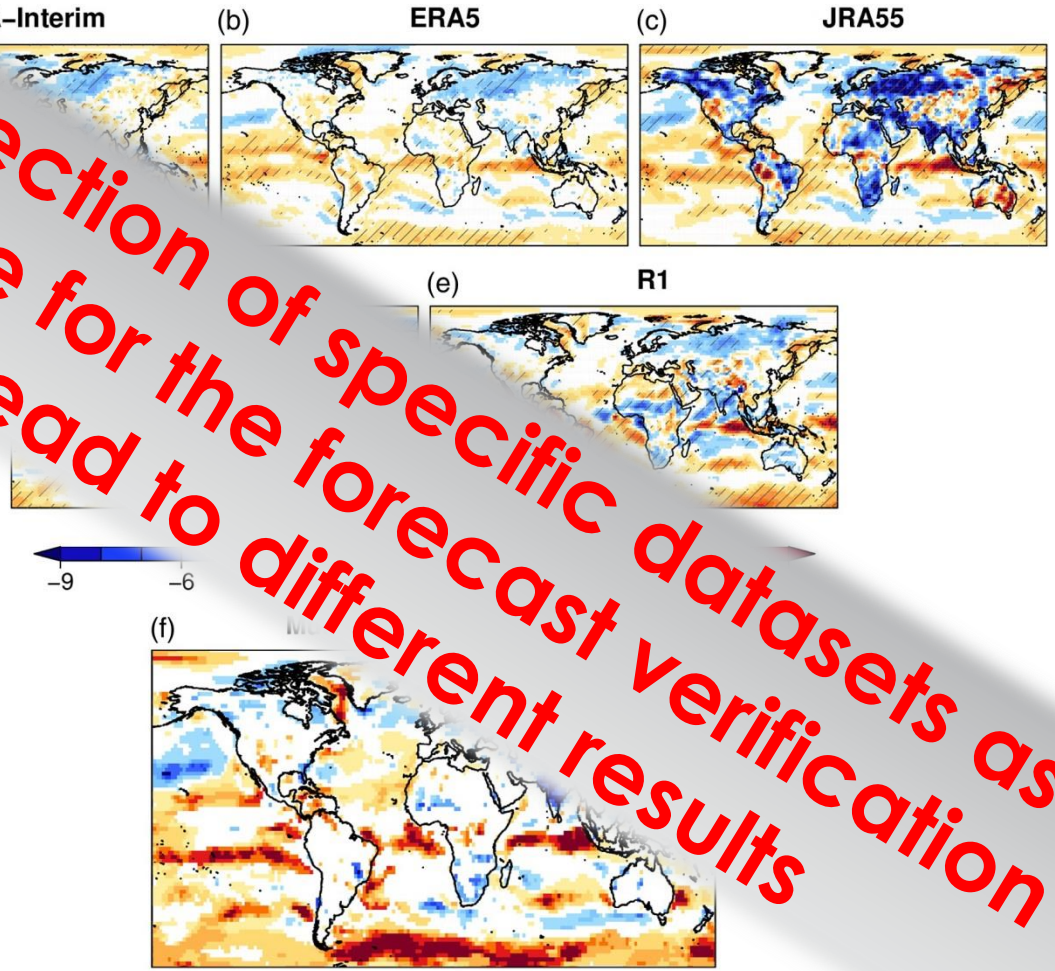
Niño3.4 SST correlation of the ensemble mean for EC-Earth3.1 (T511/ORCA025) predictions with ERAInt and GLORYS2v1 initial conditions, and BSC sea-ice reconstruction started every May over 1993-2009.



Bellprat et al. (2017, Rem. Sens. Env.)

Observational uncertainty is relevant to users

10-m wind speed trend (in percentage of the mean wind) for five reanalyses and agree[illegible] in the multi-reanalysis (MR) in DJF over 1981-2017.



The election of specific datasets as reference for the forecast verification can lead to different results

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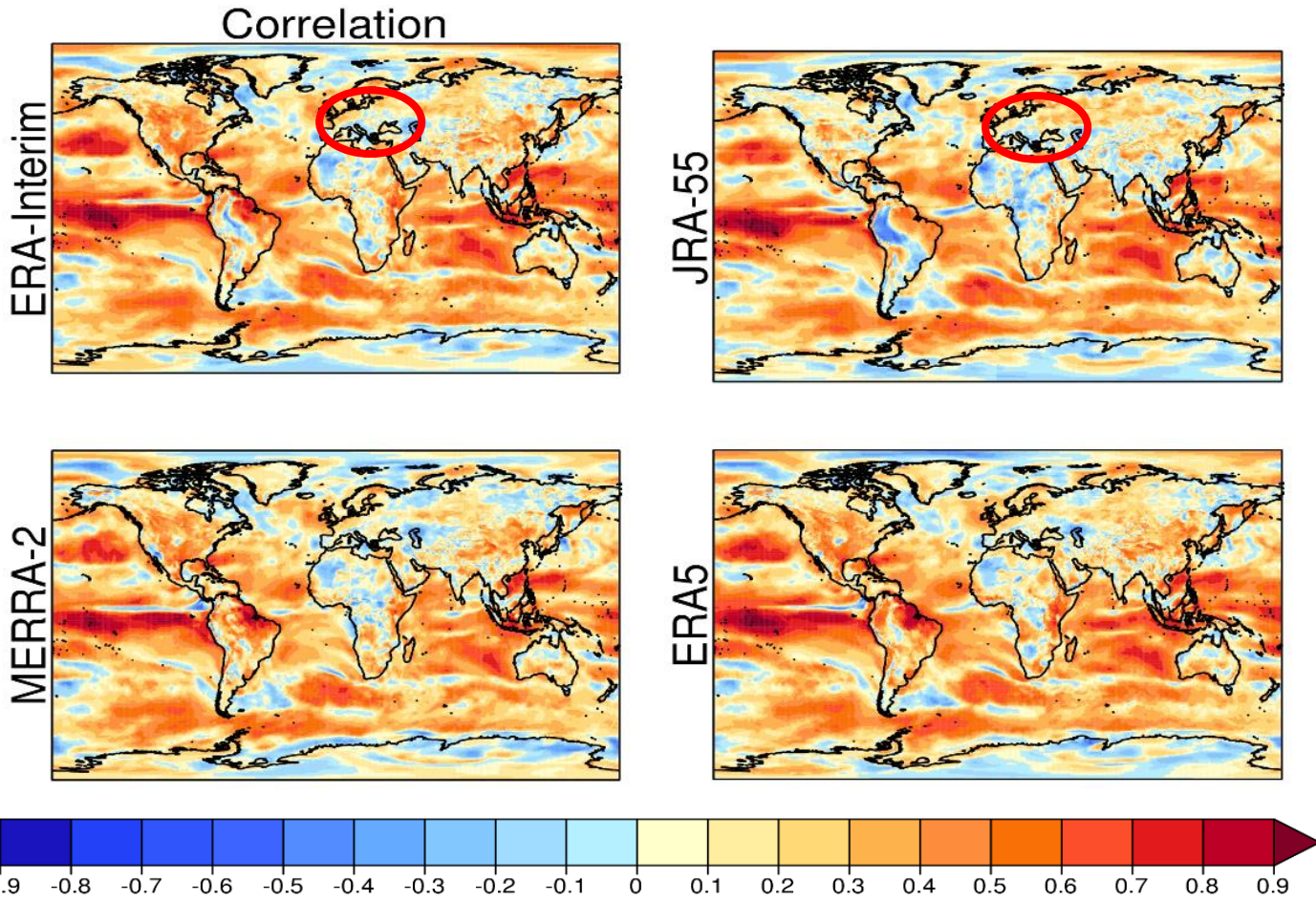
Ramon et al. (2019, QJRMS)

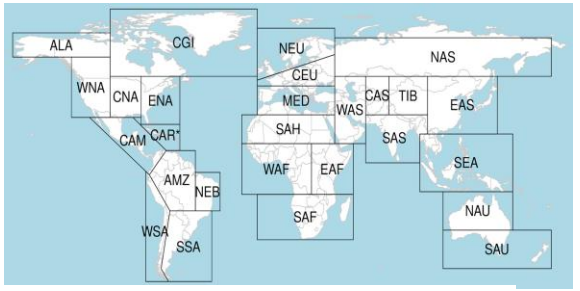
Observational uncertainty in verification

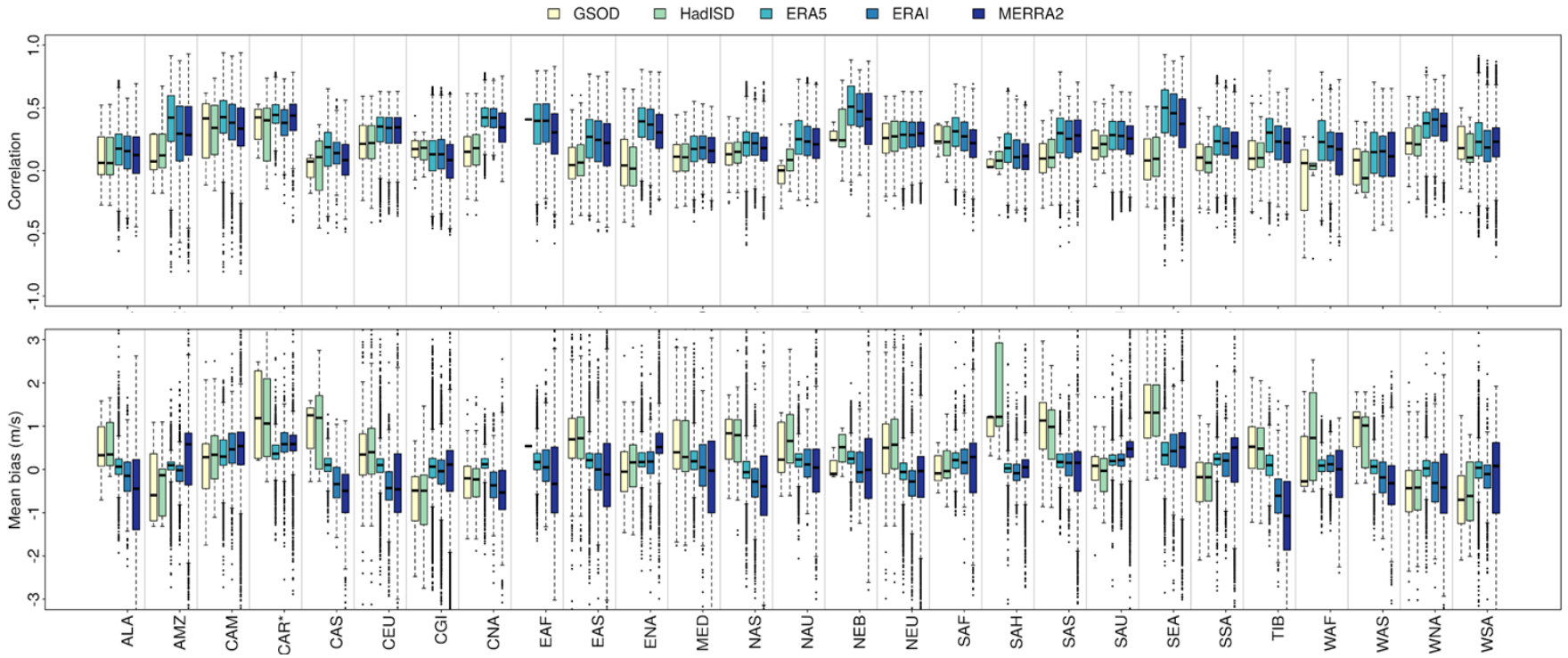
ECMWF System4
Period: 1981-2016
Season: DJF
Start date: 1st Nov
Variable: 10-m wind speed



Observational uncertainty in verification

Verification with two ground-based observational datasets and three reanalyses. The use of both types of datasets is very informative for wind energy users as they use them for the development of impact models and the long-term resource assessments.





ECMWF SEAS5 Period: 1981-2016

Season: DJF Start date: 1st Nov

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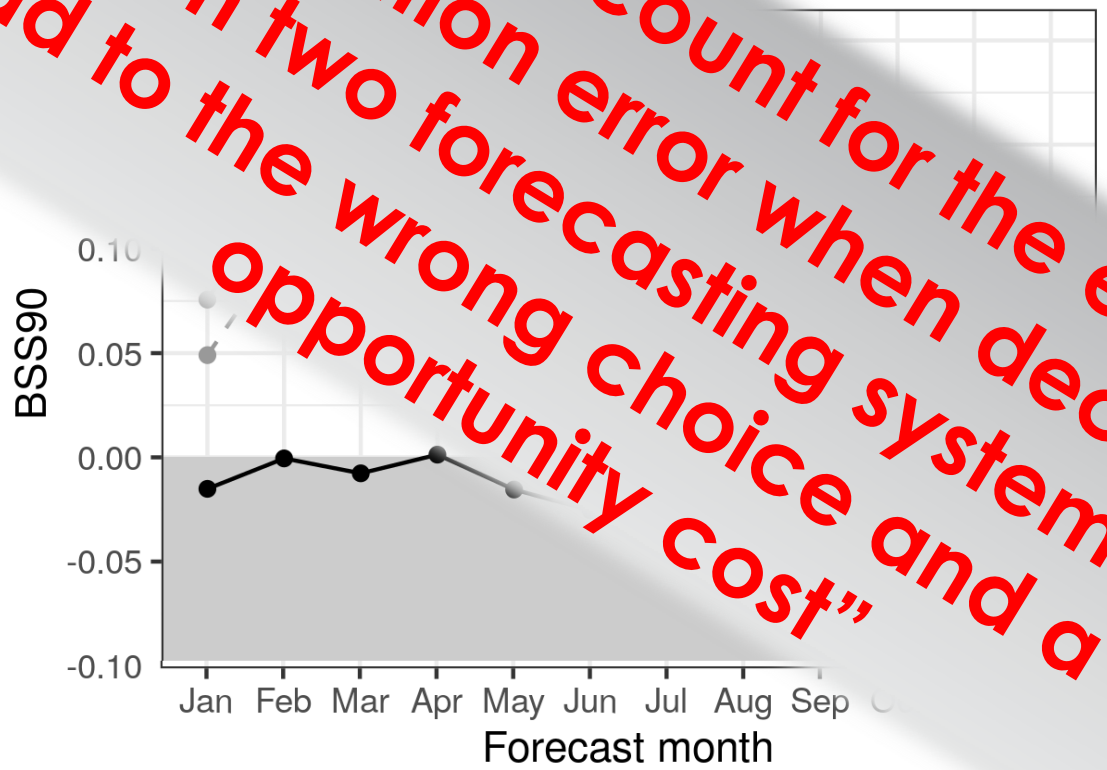
V. Torralba

Observational uncertainty in verification

Brier skill [illegible] wind speed predictions above the climatological 90th percentile [illegible] the reference forecast. Observational references from [illegible] high resolution (HRES) product with the Ferro (20[illegible]stimates in the presence of observational error



"Failing to account for the effects of observation error when deciding between two forecasting systems could lead to the wrong choice and a high opportunity cost"

ECMWF SEAS5 Period: 1981-2016
Lead time: 1 month
Global coverage, 1,542 HadISD stations, which are the truth

A non-trivial climatology definition

There is a large heterogeneity in the real-time subseasonal systems: different initialisations, hindcasts periods, etc.

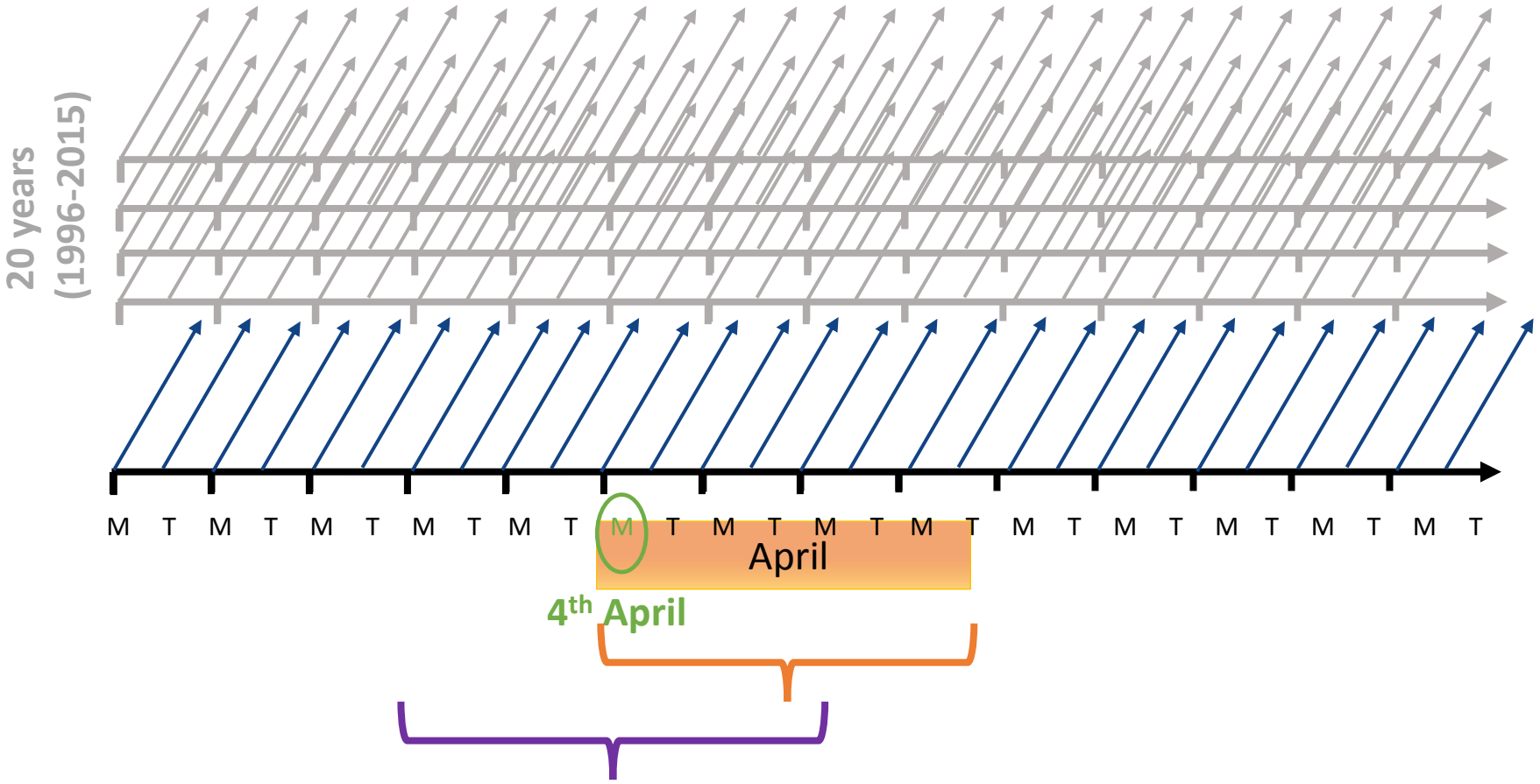
Limited samples (even in the hindcasts) lead to : lack of robustness in forecast quality estimates, definition of the climatology, bias adjustment, etc.

Forecast | Hindcasts

Status on 2020-10-27	Time range	Resolution	Ens. Size	Frequency	Re-forecasts	Rfc length	Rfc frequency	Rfc size
BoM (ammc)	d 0-62	T47L17	3*11	2/week	fixed	1981-2013	6/month	3*11
CMA (babj)	d 0-60	T266L56	4	2/week	on the fly	past 15 years	2/week	4
CNR-ISAC (isac)	d 0-32	0.75x0.56 L54	41	weekly	fixed	1981-2010	every 5 days	5
CNRM (lfpw)	d 0-47	T255L91	25	weekly	fixed	1993-2017	every 7 days	10
ECCC (cwao)	d 0-32	39 km L45	21	weekly	on the fly	1998-2017	weekly	4
ECMWF (ecmf)	d 0-46	Tco639/319 L91	51	2/week	on the fly	past 20 years	2/week	11
HMCR (rums)	d 0-61	1.1x1.4 L28	20	weekly	on the fly	1985-2010	weekly	10
JMA (rjtd)	d 0-33	TI479/TI319L100	50	weekly	fixed*	1981-2010	2/month	13
KMA (rksl)	d 0-60	N216L85	4	daily	on the fly	1991-2016	4/month	3
NCEP (kwbc)	d 0-44	T126L64	16	daily	fixed	1999-2010	daily	4
UKMO (egrr)	d 0-60	N216L85	4	daily	on the fly	1993-2016	4/month	7

Manrique-Suñén et al. (MWR, 2020)

A non-trivial climatology definition



Weekly: 1 start date, 20 years

Monthly: All start dates in a calendar month, 8/9 start dates, 20 years

Monthly running window: Running window with 4 start dates before and after the target week, 9 start dates, 20 years

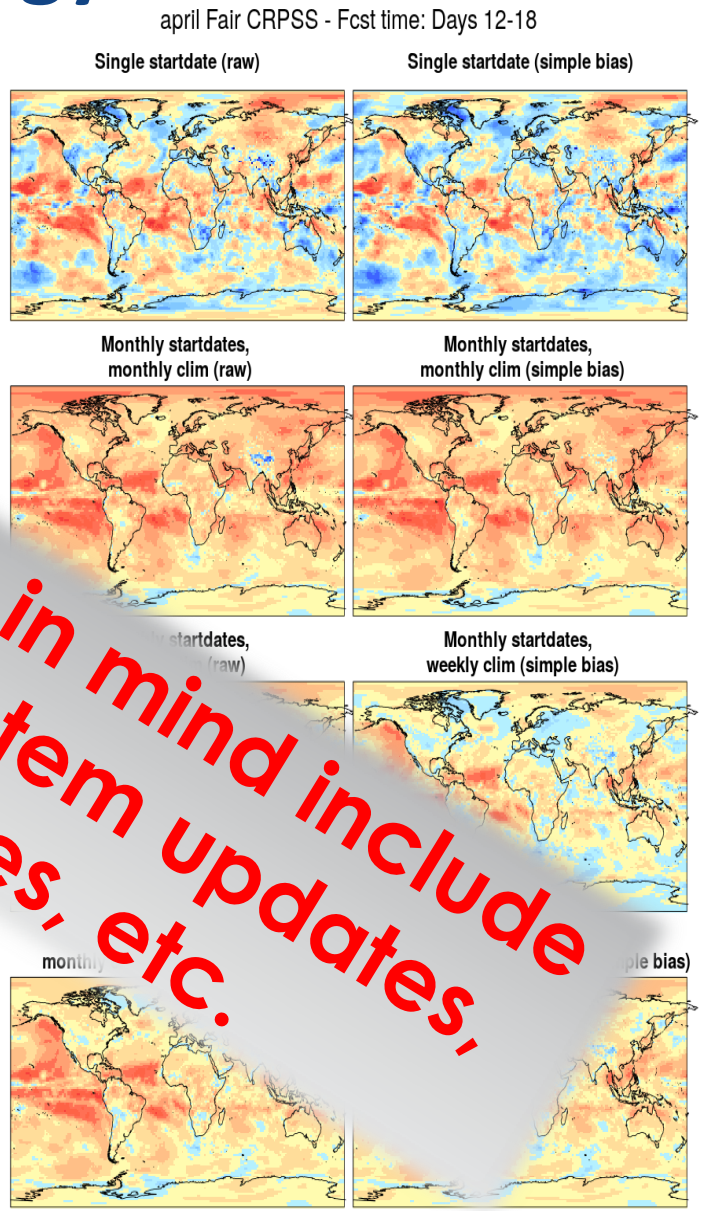
A non-trivial climatology definition

ECMWF
Period:
Reference:
Variable: 2-m tem

kly: too noisy

Monthly runni
but weekly for t
adjustment: lower skill ,
too noisy for the adjustment

Monthly running window:
more credible estimates

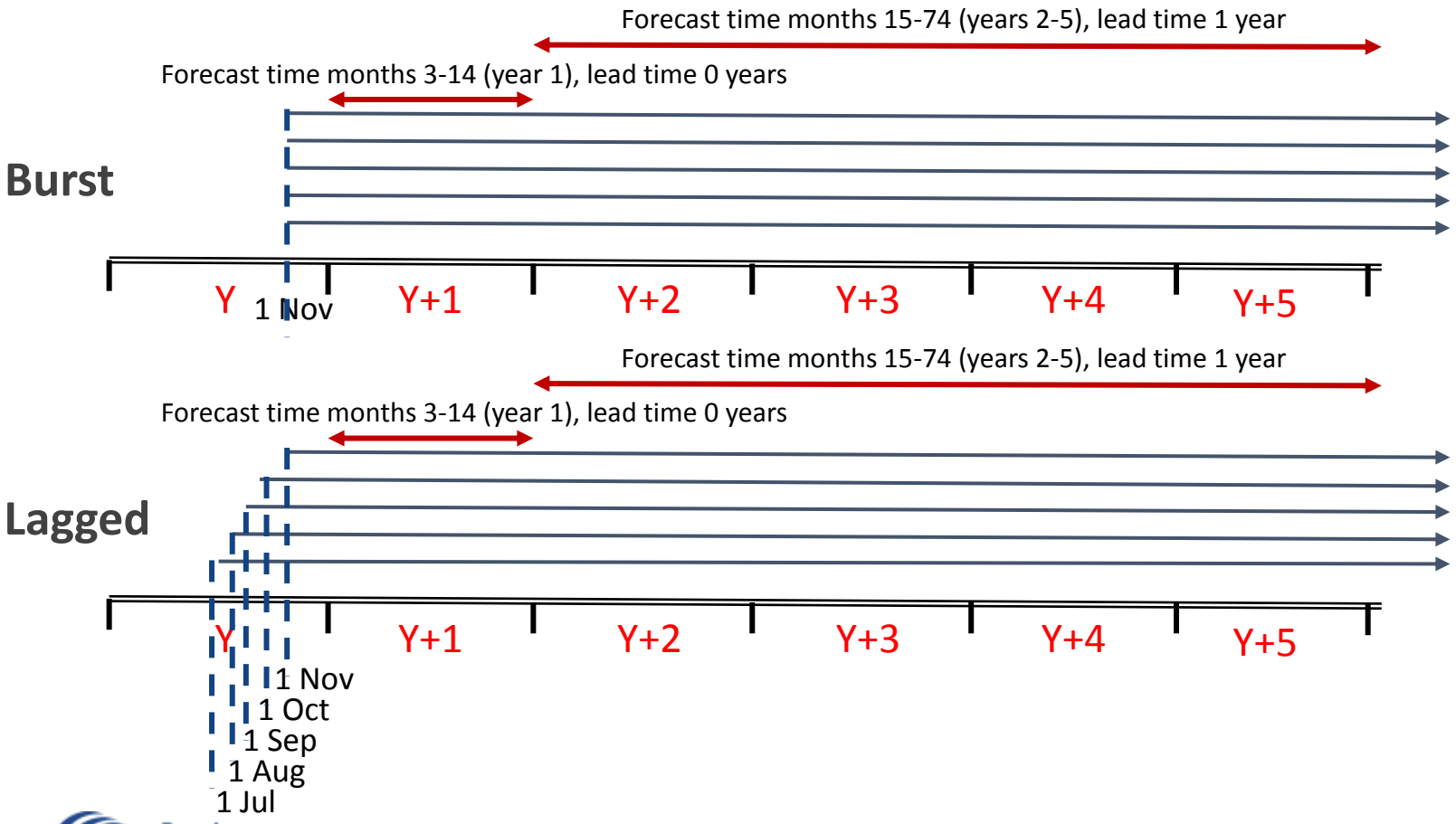


Other aspects to bear in mind include forecast timeliness, system updates, multi-model choices, etc.

Manrique-Suñén et al. (MWR, 2020)

Vocabularies

Vocabularies are part of the development of the evaluation and quality control (EQC) and a pre-requisite for the inclusion of decadal predictions in the climate data store (CDS) of the Copernicus Climate Change Service.



Evaluation and quality control

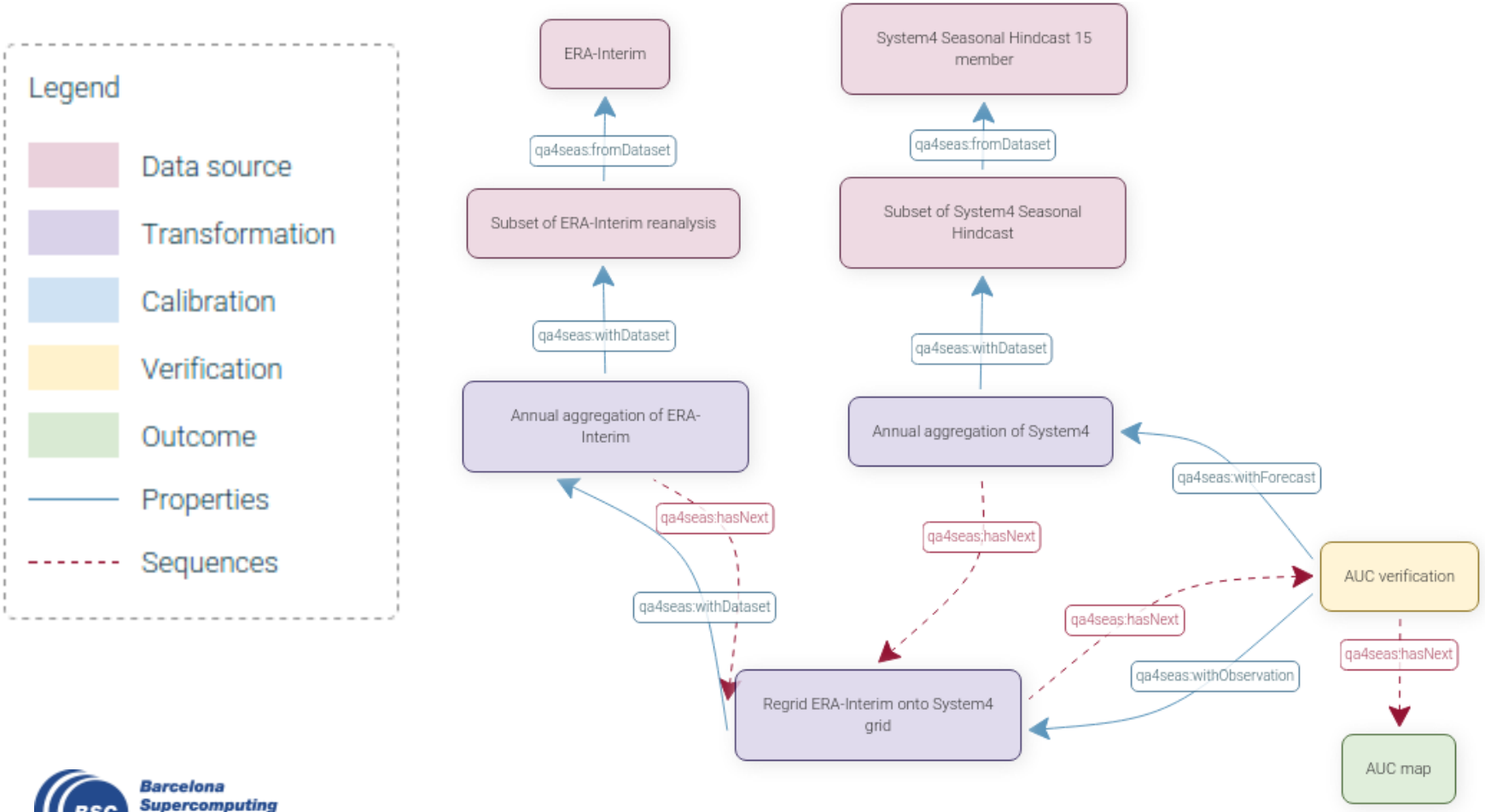
The Cop icus Climate Change Service is developing the evaluation and quality OC) function of the climate data store to:

- Pr verarching EQC service for the whole CDS
- [uality assessment of both datasets and prod

Forecast quality assessment is a part of the evaluation and quality control function

How traceable are both products and skill?

Generalised metadata provision and workflow provenance is required to ensure a minimum quality of the forecast-based climate information.



How traceable are both products and skill?

Well-documented packages safely developed are indispensable in an operational context.

The Comprehensive R Archive Network

- package shared in ftp and web servers around the world
- common standards for R users
  - easy installation
  - typical R documentation
  - typical warnings and error messages
- checks
  - to work on different OS
  - structure and documentation
  - package size
  - installation time
  - ...

Profiling tools

- ProfVis and Rprof
- scalability and parallelisation

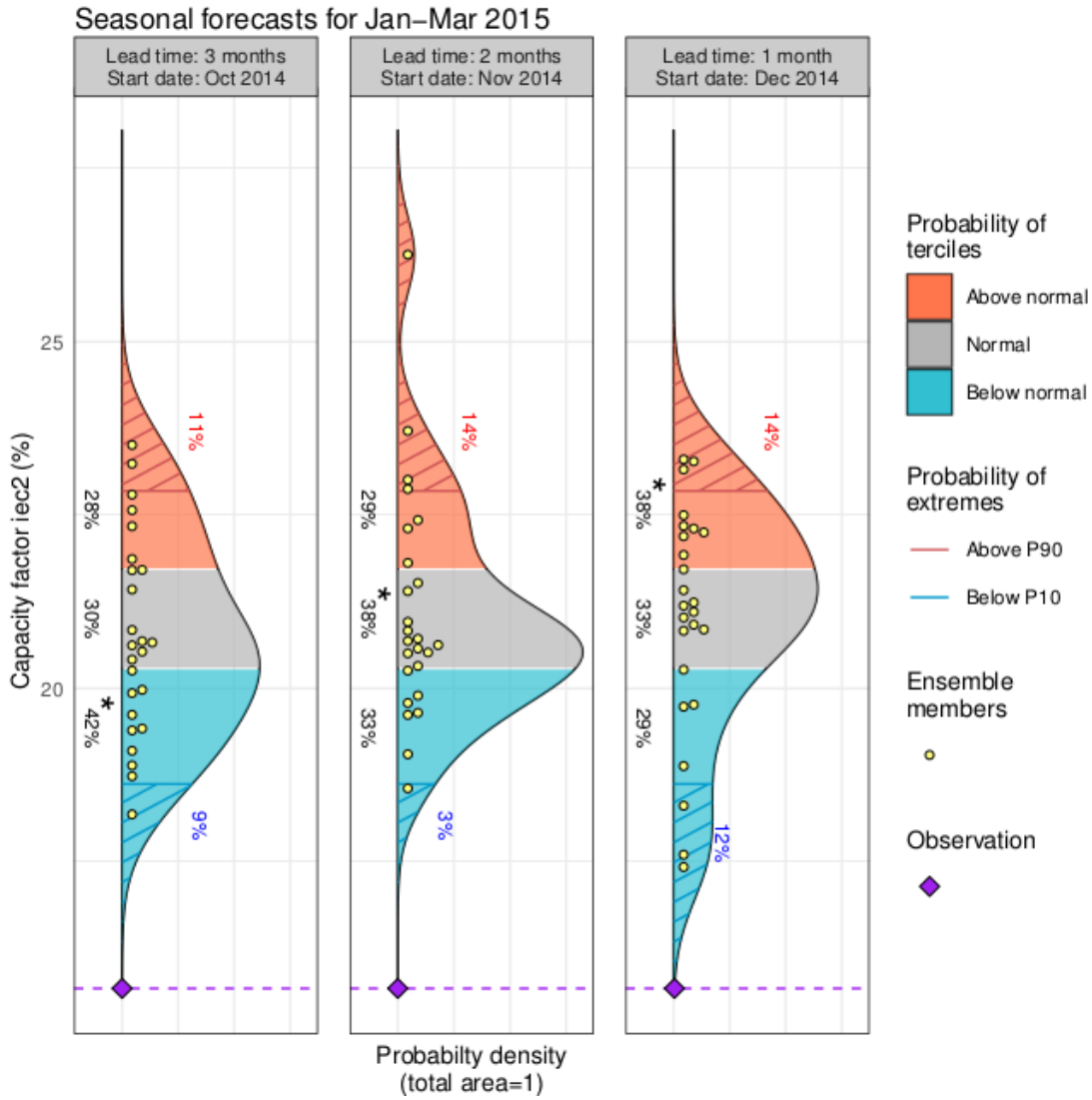
Version control system

- track changes
- issues discussion
- tag versions
- branching strategy
- unit testing
- continuous integration

Contributions and improvements

- open a new issue
- work in a new branch
- follow function standards
- review process
- merge into the master branch
- plan next release

Seasonal prediction of wind capacity factor



Seasonal predictions of DJF capacity factor over North America (124-95°W, 26-44°N) starting on the first of October, November and December for the first trimester of 2015, ECMWF SEAS5, reanalysis: ERA-Interim, hindcasts over 1993-2015.

	Oct	Nov	Dec
RPSS	0.23	0.25	0.24
BS P10	–0.18	–0.23	–0.16
BS P90	0.06	0	0.03
CRPSS	0.11	0.08	0.08
EnsCorr	0.5	0.45	0.42

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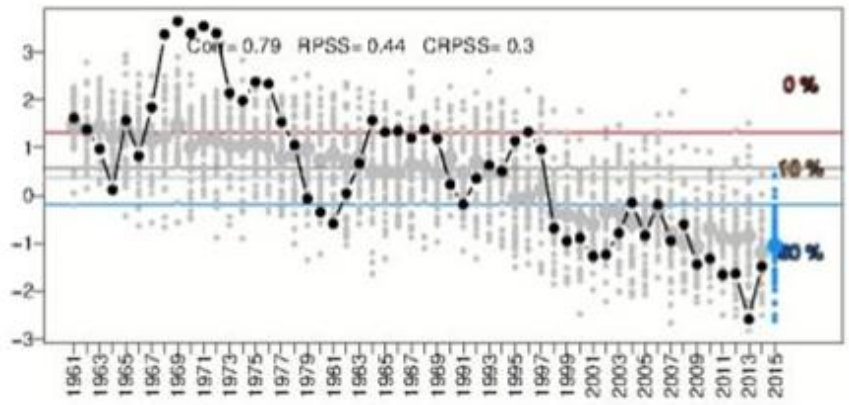
L. Lledó

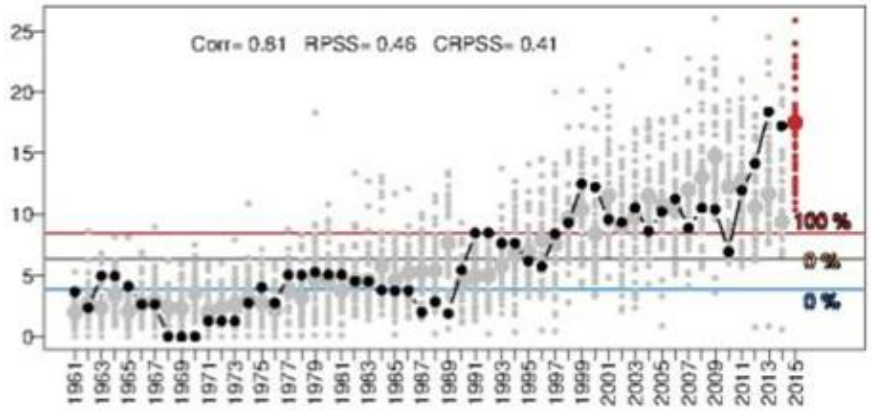
Decadal prediction of crop yield indices

WMO recognised global producing centres of decadal predictions contribute with the definition of standards for decadal predictions data and products, while C3S promotes the evaluation of the European multi-model and the illustration of the decadal prediction use in, among other sectors, the agricultural sector.

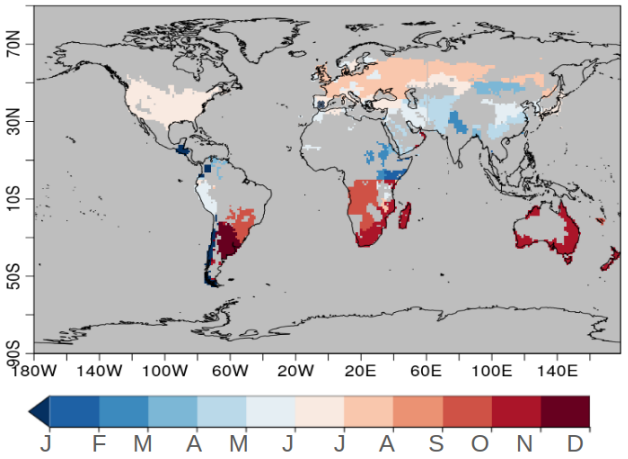
SPEI6



HMDI3



Wheat harvesting month



Indicators:

- Drought: Standardized Precipitation Evapotranspiration Index (SPEI6)
- Heat Stress: Heat Magnitude Day Index (HMDI3)

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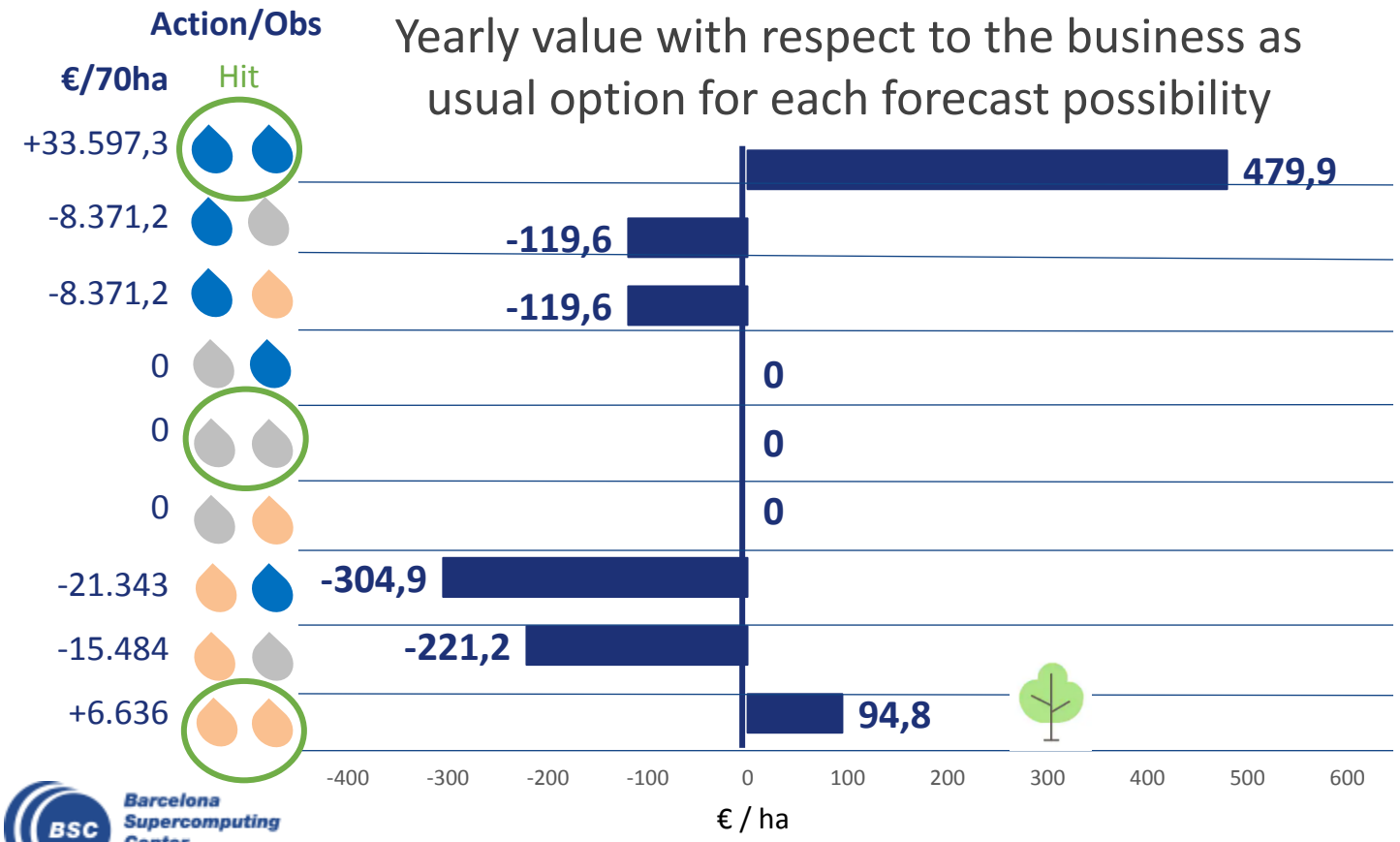
Copernicus Europe's eyes on Earth

Climate Change Service

B. Solaraju

User value: wine making

A wine producer is using seasonal forecasts of spring rain (provided in January) to make decisions about with the objective of maximising the payoff of the production of a 70 ha lot. Forecast value is user specific and is the basis for the user-oriented forecast application.



Spring Rain (SP)	
Above Average	
Normal	
Below Average	
Plant Protection Costs	C_{pp}
Canopy Management Costs	C_{cm}
Yields	Y

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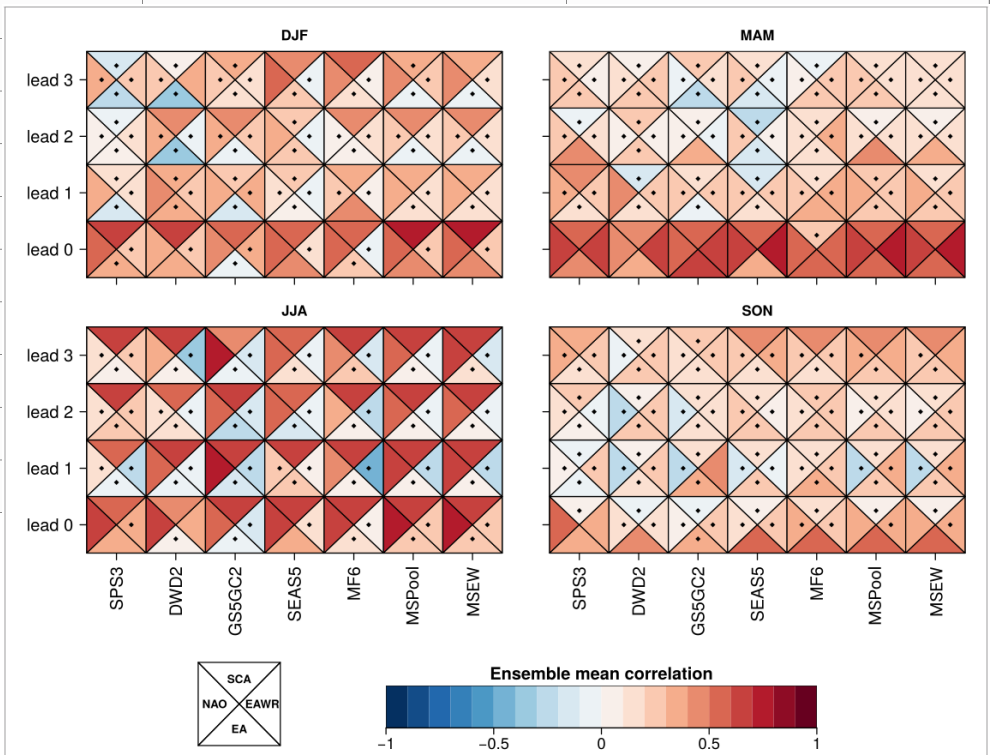
Do not underestimate the power of an image

Scorecards and comprehensive displays of forecast quality measures become essential in a user-oriented service.

Mean sea level pressure	CMCC System 3	Mean Bias (Pa)						Correlation						FCRPSS						FRPSS					
		Forecast month						Forecast month						Forecast month						Forecast month					
Start date	Region	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6	1	2	3	4	5	6
January	Tropics (30S-30N)	3.01	6.95	0.12	9.44	3.09	9.19	0.74	0.55	0.35	0.40	0.25	0.22	0.29	0.12	0.00	0.04	0.05	0.04	0.35	0.16	0.03	0.08	0.02	0.01
	Extra-tropical NH (30N-90N)	3.51	8.24	7.80	0.18	9.04	0.89	0.58	0.24	0.08	0.17	0.12	0.12	0.12	0.04	0.09	0.06	0.06	0.07	0.18	0.01	0.05	0.02	0.02	0.02
	Extra-tropical SH (30S-90S)	2.49	5.76	4.56	0.33	0.30		0.65	0.10	0.09	0.05	0.04	0.03	0.20	0.08	0.07	0.06	0.09	0.07	0.24	0.07	0.03	0.02	0.07	0.05
February	Tropics (30S-30N)	0.75	7.35	5.48	5.40	2.89	1.10	0.73	0.45	0.45	0.23	0.30	0.40	0.25	0.05	0.07	0.02	0.00	0.05	0.35	0.10	0.10	0.01	0.03	0.09
	Extra-tropical NH (30N-90N)	1.39	0.36	0.15	7.83	7.57	3.42	0.62	0.27	0.20	0.13	0.09	0.13	0.12	0.03	0.04	0.06	0.07	0.05	0.20	0.00	0.01	0.02	0.03	0.02
	Extra-tropical SH (30S-90S)	2.94	5.92	4.22	0.10			0.52	0.19	0.09	0.01	0.09	0.09	0.05	0.03	0.06	0.07	0.06	0.05	0.12	0.00	0.02	0.04	0.05	0.02
March	Tropics (30S-30N)	2.85	5.14	8.88	0.47	4.40	6.42	0.68	0.47	0.38	0.38	0.39	0.40	0.20	0.08	0.00	0.02	0.04	0.03	0.26	0.12	0.03	0.05	0.07	0.07
	Extra-tropical NH (30N-90N)	0.15	0.22	2.15	7.65	2.38	7.29	0.63	0.26	0.18	0.08														
	Extra-tropical SH (30S-90S)	8.89	5.32	4.72				0.42	0.05	0.05	0.05														
April	Tropics (30S-30N)	7.43	3.03	8.44	3.46	6.52	1.74	0.64	0.42	0.40	0.46														
	Extra-tropical NH (30N-90N)	7.72	0.31	8.27	3.88	5.71	7.11	0.58	0.19	0.13	0.24														
	Extra-tropical SH (30S-90S)	0.64	4.53					0.44	0.09	0.23	0.22														
May	Tropics (30S-30N)	1.23	9.14	9.02	8.68	1.25	5.28	0.64	0.43	0.44	0.48														
	Extra-tropical NH (30N-90N)	0.00	1.53	4.06	9.49	5.01	0.82	0.49	0.11	0.10	0.10														
	Extra-tropical SH (30S-90S)	0.22						0.57	0.17	0.14	0.01														
June	Tropics (30S-30N)	8.87	9.94	4.91	1.01	4.13	1.91	0.71	0.47	0.48	0.44														
	Extra-tropical NH (30N-90N)	7.68	2.96	2.72	8.85	0.14	1.90	0.51	0.18	0.13	0.06														
	Extra-tropical SH (30S-90S)	0.75						0.52	0.25	0.16	0.08														
July	Tropics (30S-30N)	1.89	2.33	1.36	1.08	4.71	5.98	0.68	0.53	0.47	0.44														
	Extra-tropical NH (30N-90N)	0.92	6.15	7.00	8.43	7.84	1.14	0.50	0.17	0.15	0.04														
	Extra-tropical SH (30S-90S)	5.56						0.52	0.03	0.14	0.22														
August	Tropics (30S-30N)	7.68	5.85	6.48	3.56	0.59	2.39	0.72	0.45	0.47	0.42														
	Extra-tropical NH (30N-90N)	1.29	4.53	0.54	7.49	9.71	1.02	0.43	0.06	0.07	0.22														
	Extra-tropical SH (30S-90S)	2.83						0.53	0.20	0.24	0.37														
September	Tropics (30S-30N)	4.29	0.21	3.60	4.66	8.11	0.61	0.70	0.52	0.46	0.51														
	Extra-tropical NH (30N-90N)	1.49	3.01	9.99	1.64	-5.07	-0.76	0.43	0.07	0.10	0.07														
	Extra-tropical SH (30S-90S)	3.53						0.57	0.30	0.38	0.10														
October	Tropics (30S-30N)	1.84	3.49	7.38	6.09	1.12	6.22	0.66	0.45	0.50	0.52														
	Extra-tropical NH (30N-90N)	1.42	2.84	5.87	-0.87	-3.59	4.06	0.38	0.08	0.14	0.07														
	Extra-tropical SH (30S-90S)	3.69						0.57	0.24	0.08	0.12														
November	Tropics (30S-30N)	8.38	2.67	6.00	5.52	6.69	0.63	0.70	0.53	0.53	0.47														
	Extra-tropical NH (30N-90N)	0.69	9.69	1.99	1.32	3.78	9.93	0.53	0.17	0.07	0.17														
	Extra-tropical SH (30S-90S)	5.93	3.02	3.74	1.80	3.71	1.80	0.64	0.23	0.19	0.13														
December	Tropics (30S-30N)	0.86	5.61	9.57	3.90	7.20	1.17	0.76	0.57	0.48	0.38														
	Extra-tropical NH (30N-90N)	4.70	4.38	1.69	2.43	5.19	5.56	0.60	0.15	0.15	0.04														
	Extra-tropical SH (30S-90S)	5.82	5.33	2.46	3.17	0.02	2	0.64	0.22	0.16	0.03														

Forecast quality measures for the CMCC-CM3 operational seasonal forecast system from C3S.

N. Cortesi

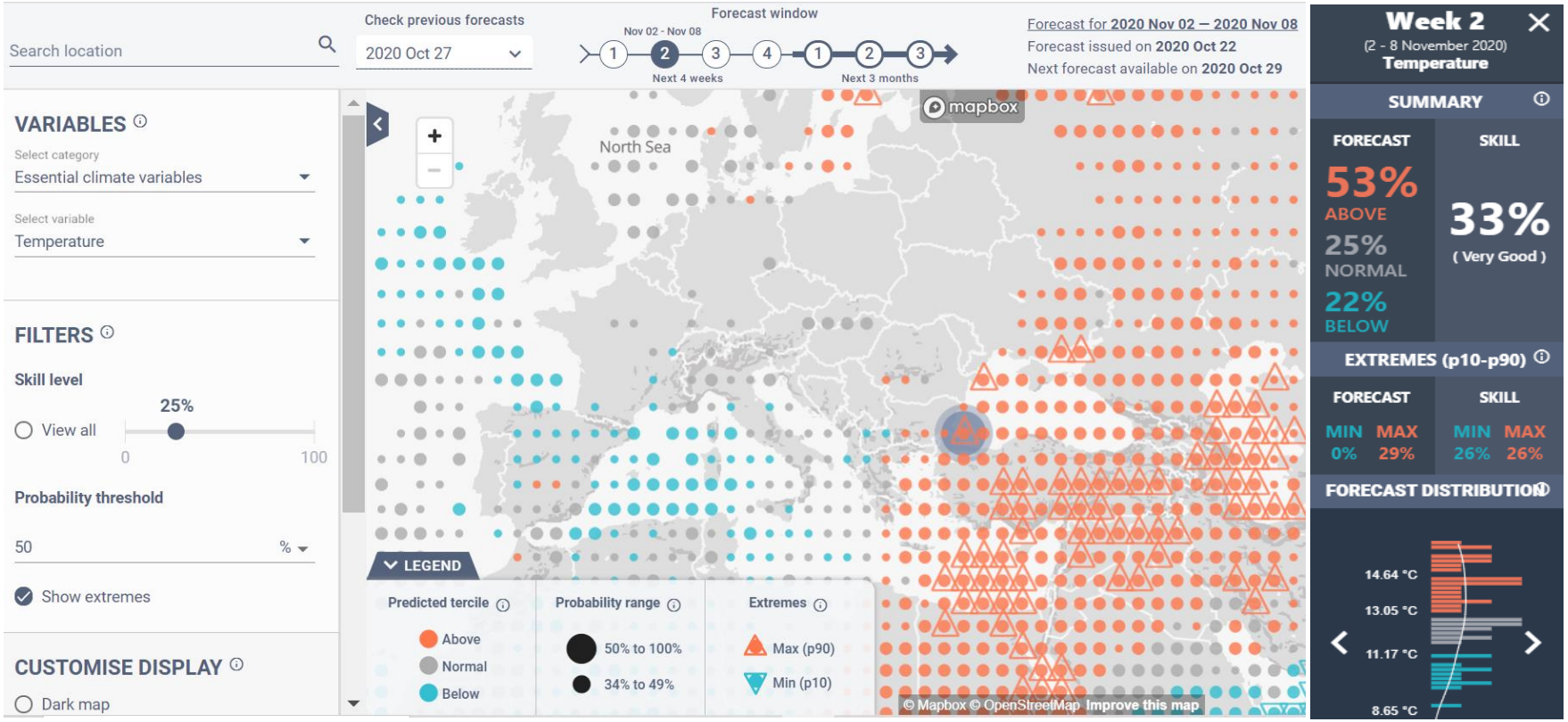


Correlation for C3S seasonal forecasts of North Atlantic modes of variability.

Lledó et al. (2020, ERL)

Prototypical climate services for energy

S2S4E is developing a decision support tool for the renewable energy sector based on Copernicus climate forecasts, S2S, and NCEP operational predictions co-designed with the industry for periodic updates on the state of relevant climate variables.



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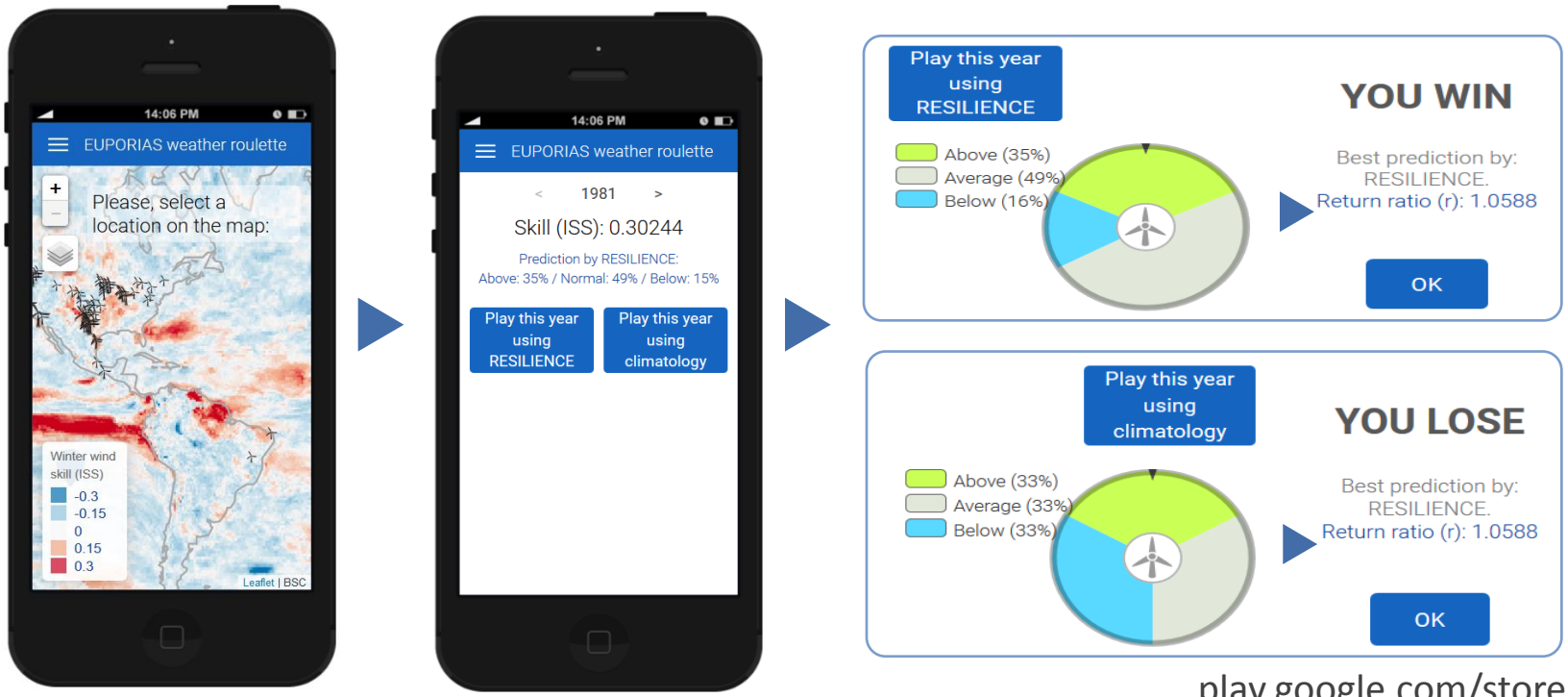
Subseasonal-to-Seasonal S2S Prediction Project

The communication challenge

Gamification is useful to illustrate the challenges of using and the value of seasonal climate predictions addressed to the wind energy sector:

- Play against a reference taken from climatological frequencies.
- The bets are proportional to the predicted probabilities.
- The amount invested in the observed category is multiplied by 3.



play.google.com/store/apps

demo.predictia.es/roulette-app/mobile.html

Terrado et al. (2019, BAMS)

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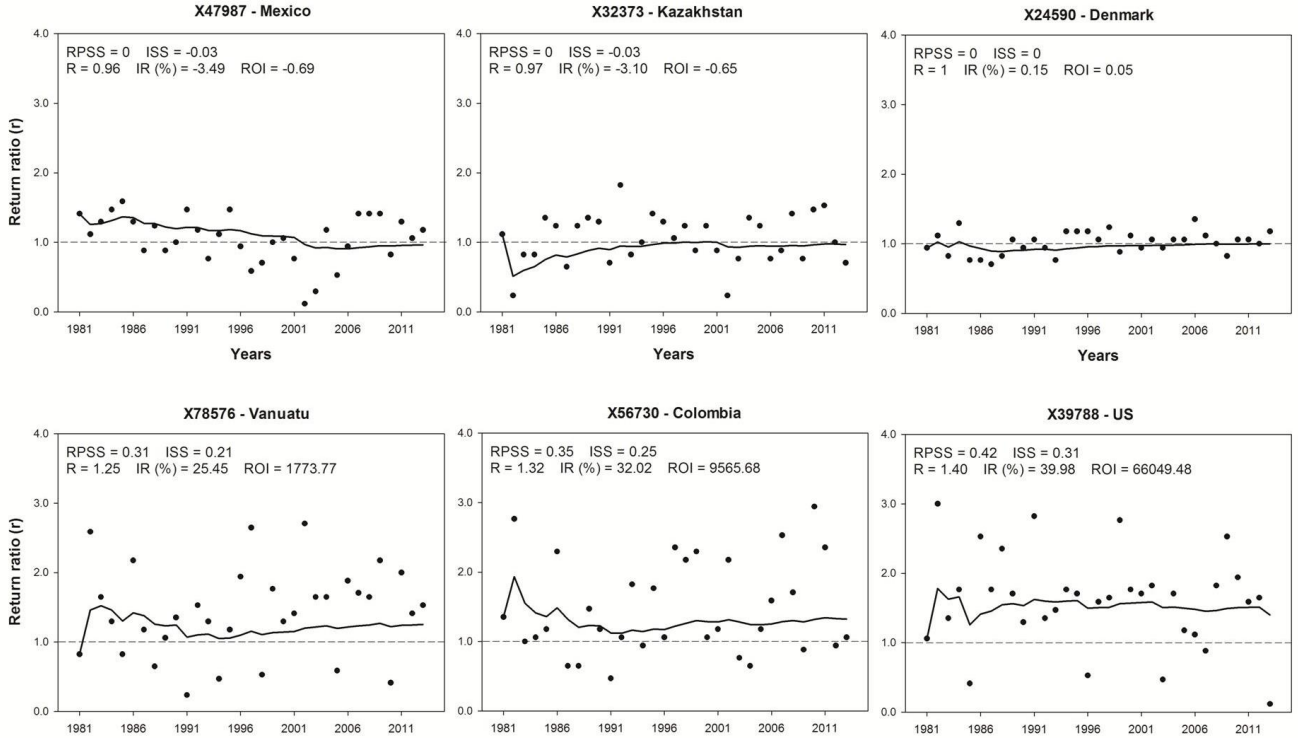
Illustrating the value of predictions

Examples of return ratio for 33 betting runs for different points where wind power plants are installed:

- Top row cases with RPSS=0, but ignorance skill score negative or zero.
- Bottom row cases with RPSS>0.
- Line for the geometric average of return ratios (interest rate).



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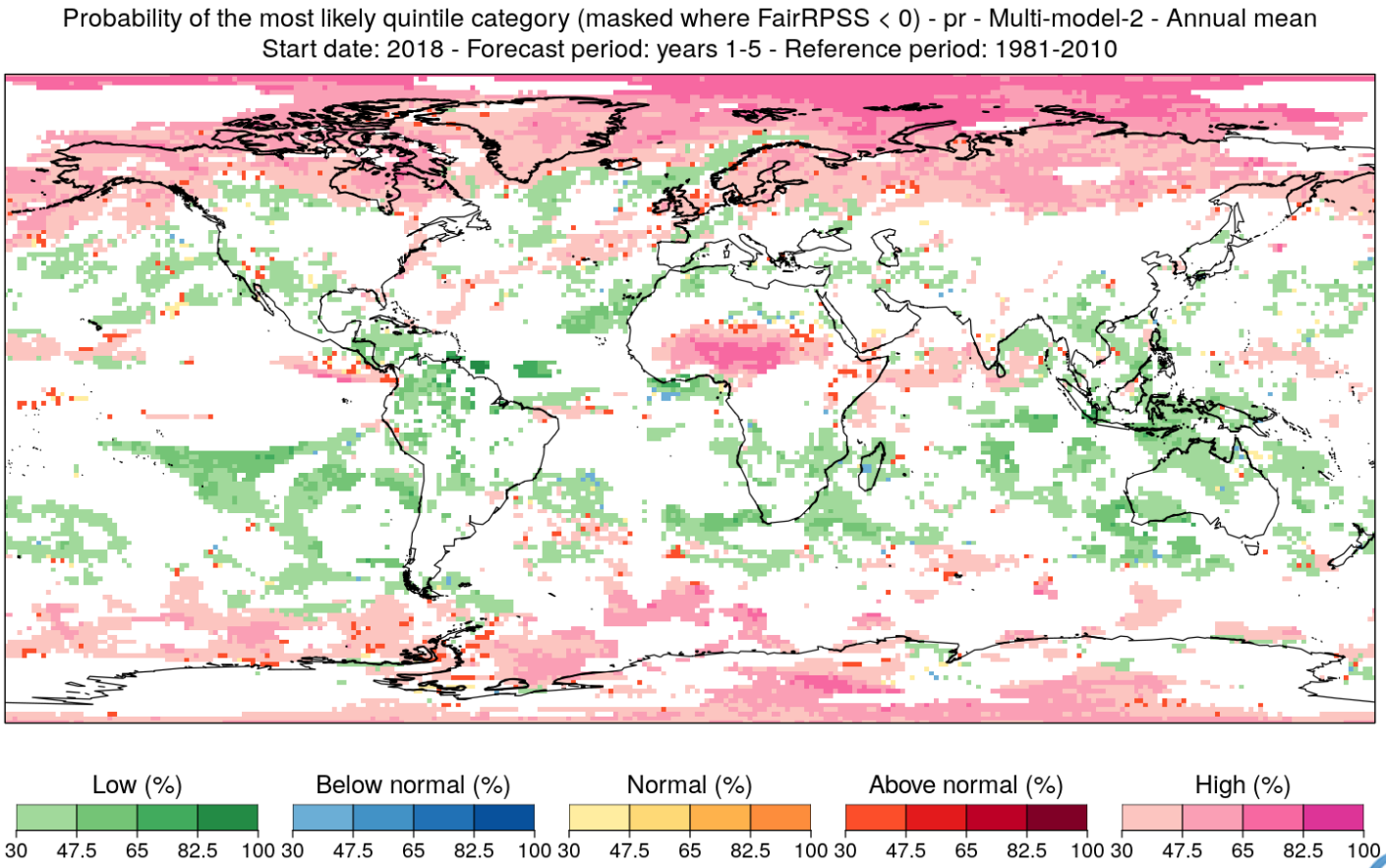
Multi-model climate predictions

Multi-model forecasts have been considered as an acceptable way to synthesise forecast information.

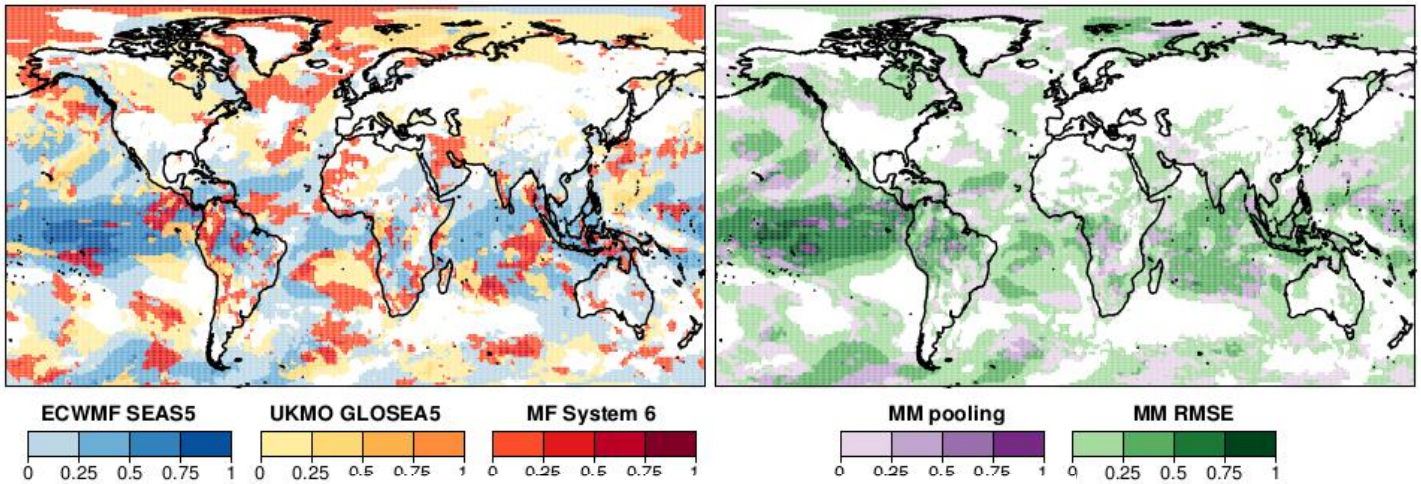
Decadal predictions of precipitation for forecast years 1-5 started near the end of 2018.

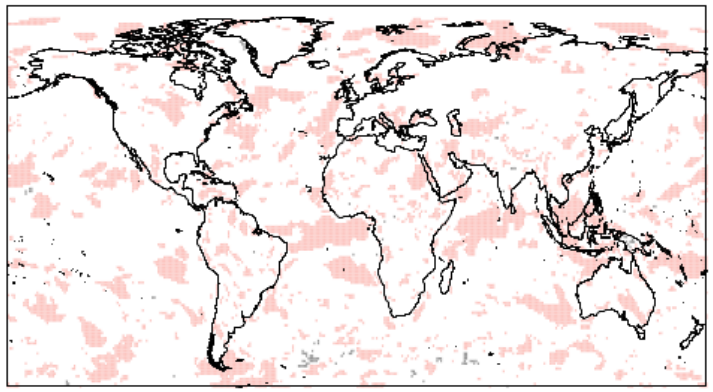


Multi-model climate prediction

However, users do not always understand why multi-model is the preferable option.

CRPSS of DJF two-metre temperature for C3S forecasts initialized in November, all systems bias adjusted (MVA) compared to a simple and weighted multi-model (as inverse function of RMSE). Bottom gain of the best multi-model with respect to the best single system. Verified against ERA Interim for 1993-2015.





The non-exhaustive list of relevant elements

- Observational uncertainty: comparison between reanalyses in a forecast verification context.
- Definition of standard procedures: standards are less common than one would expect.
- Traceability and quality control: quality control and reproducibility of data and products is increasingly important in the research community, but its operational aspects are not solved yet.
- User indicators: indicators do not have the same level of skill as the meteorological variables.
- Interpretation and communication: users are often not experts, and even when they are it is easy to misunderstand the existing information. Communication is a challenge
- Synthesis and narratives: how to deal with multiple lines of evidence in the message constructions.