#### Reading, 5 September 2019



Barcelona Supercomputing Center Centro Nacional de Supercomputación

EXCELENCIA SEVERO OCHOA

Sub-seasonal and Seasonal Climate Forecast Applications

Francisco J. Doblas-Reyes

with contributions from Joaquín Bedia, Nicola Cortesi, Carlo Lacagnina, Llorenç Lledó, Jaume Ramón, Daniel San Martín, Marta Terrado, Verónica Torralba, Marco Turco, Ilaria Vigo and many more

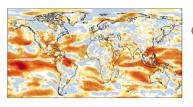




# **Some definitions**



 Climate data: numerical values representing ECVs and associated magnitudes; includes metadata



Climate information: the result of processing climate data according to the available climate knowledge



Climate message: climate information put in context and generated in a co-production process

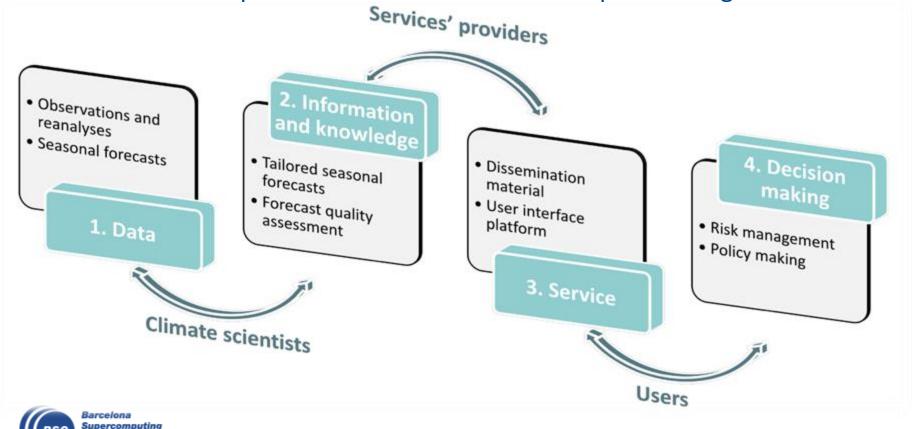


 Climate service: ensemble of processes through which a climate message is generated



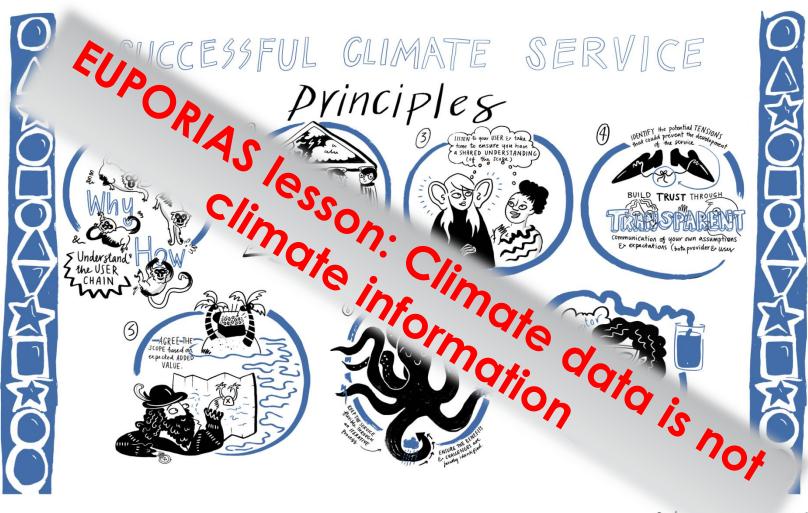
# The research-provider-service paradigm

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 illustrate. This leaves a space for transdisciplinary research. This chain should not preclude basic research to take place though.



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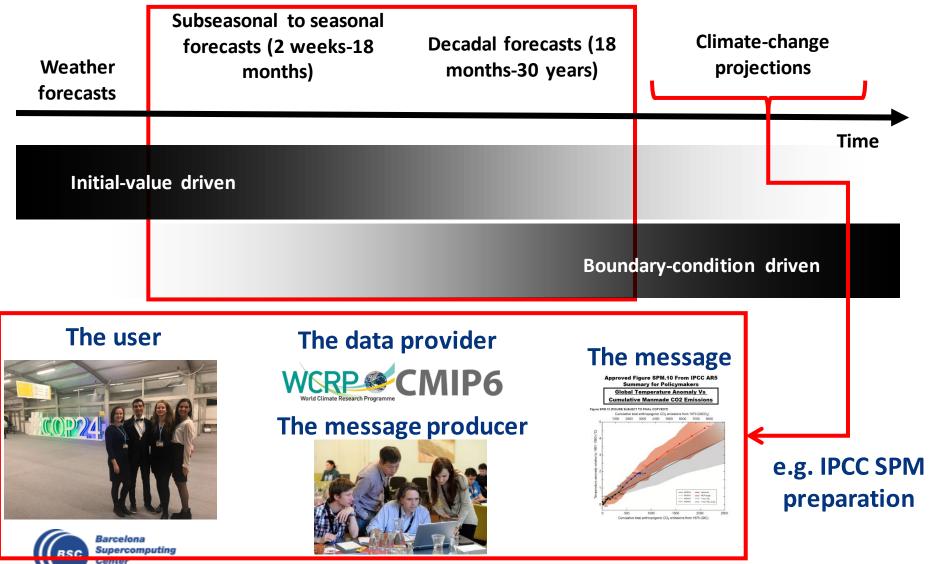
# The research-provider-service paradigm





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# **Climate time scales**



Adapted from Meehl et al. (2009)

# **Barriers to use climate forecasts in applications**

**Some limitations** for the use of climate forecasts in different socio-economic sectors.



BSC Barcelona Supercomputing Center Centro Nacional de Supercompulación **Possible solution:** to adapt the climate forecasts into a message to be integrated in decision-making.

## **Climate services**

- Goal: the development and incorporation of climate messages based on forecasts for planning, policy-making and practice at the global, regional and national scale.
- Implementation method: coproduction and co-design.

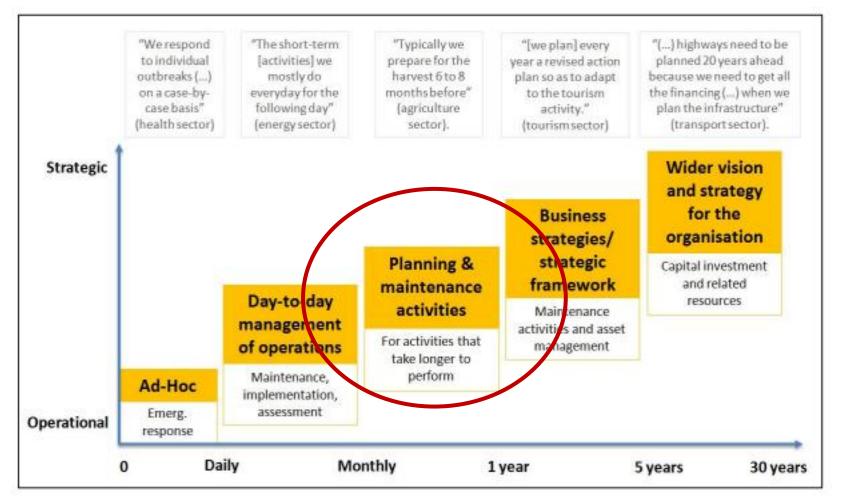
# Some elements to produce climate messages

- User engagement: raise awareness, identification of requirements, indicators, roadmap for co-production, message format, funding model
- Climate data: observations, reanalyses, forecasts and hindcasts and measures of uncertainty for all of them
- Forecast product: unequivocal definition, adequate verification
- Forecast quality: verification (with uncertainty estimates), standards and data correctness, documentation, guidance, independent assessment
- Traceability: accessibility of the methodology, standards, provenance
- Message synthesis: elaboration of the message, consider difficulties for interpretation, including its periodicity, using multiple lines of evidence
- Training: gamification, workshops, case studies, webinars
- The prediction has value if it can be used to build a climate message that helps the user to obtain some kind of benefit from the decisions she has to make. Let's leave it to them.



# **User engagement: problem definition**

## Users of climate predictions are involved in more strategic decisions than those of weather forecasts





#### Dessai and Bruno-Soares (2013, SRI Paper 62)

# An example: Users in the energy sector

Long-term user engagement has allowed identifying a number of decisions that could benefit from climate predictions

#### **Post-construction decisions**

Energy producers: Resource management strategies
Energy traders: Resource effects on markets
Plant operators: Planning for maintenance works, especially offshore wind O&M
Plant investors: anticipate cash flow, optimize return on investment

## **Mid-term planning**

**Grid operators:** Anticipate hotter/colder seasons to schedule power plants to reinforce supply. **Energy traders:** Anticipate energy prices.



# Sometimes, the user wants to understand



#### ENERGY CASE STUDY

Effects of Arctic sea ice on energy production in mid-latitudes



#### CHAIN OF EVENTS

1. Historical low sea ice concentration in the Barents and Kara (BK) seas.



Barcelona Supercomputing Center Centro Nacional de Supercomputación CASE STUDY 1



#### **BASIC FACTS**

S2S4E

- Area: Europe
- Season: winte
- Year: 2018
- Forecast range: sub-seasonal
- Main Interest: electricity demand
- Forecast variables: temperature and electricit demand

This factsheet is based on S2 deliverable 4.1. To access the report, please visit s2s4e.eu.

#### WHAT happened

An unanticipated cold spell resulted in below average weekly temperatures and triggered an increase in power demand for heating.

CASE STUDY #7

#### WHERE it affected

The cold spell affected mostly eastern and central Europe.

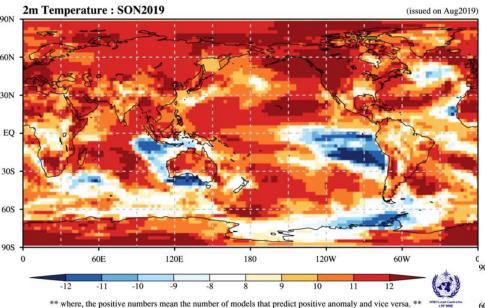
#### WHEN it occurred

This case study analyses the extreme cold temperatures that occurred from February 27th to March 5th, 2018.

# Other times, they look for climate forecasts

#### **Consistency Map**

CPTEC, ECMWF, Exeter, Melbourne, Montreal, Moscow, Offenbach, Pretoria, Seoul, Tokyo, Toulouse, Washington



## But some elements are missing:

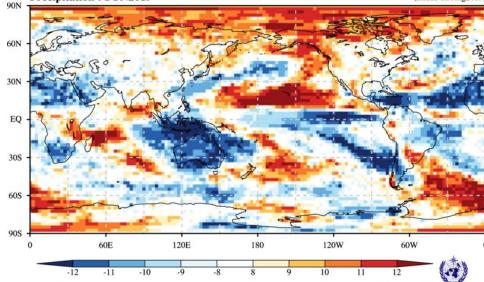
- Quality assurance
- Traceability
- Interpretation

#### **Consistency Map**

CPTEC, ECMWF, Exeter, Melbourne, Montreal, Moscow, Offenbach, Pretoria, Seoul, Tokyo, Toulouse, Washington

(issued on Aug2019)

#### Precipitation : SON2019



\*\* where, the positive numbers mean the number of models that predict positive anomaly and vice versa. \*\*



# **Examples of dealing with missing elements**

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

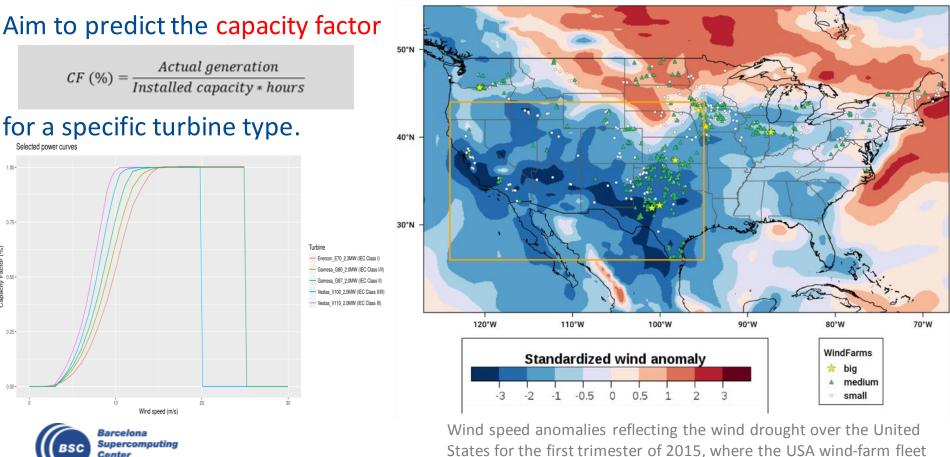


User engagement and forecast quality

L. Lledó

# Some interactions are motivated by an event

During the first quarter of 2015 the United States experienced a widespread and extended episode of low surface wind speeds. Some wind farms did not generate enough cash for their steady payments, decreasing the value of wind farm assets.

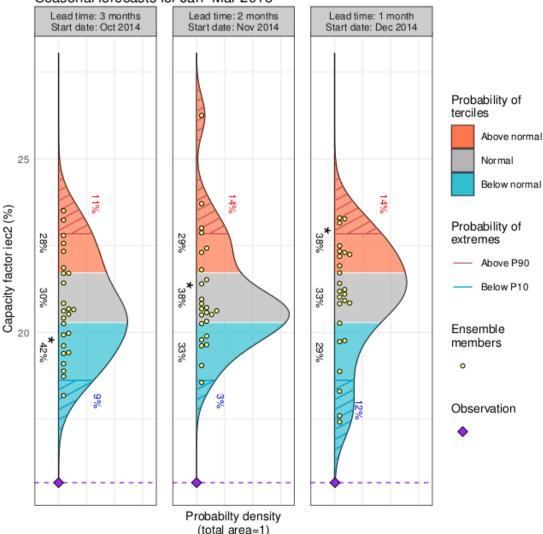


is also shown (Lledó et al., JGR 2018)

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# User engagement and forecast quality Capacity factor forecasts

#### Seasonal forecasts for Jan-Mar 2015



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

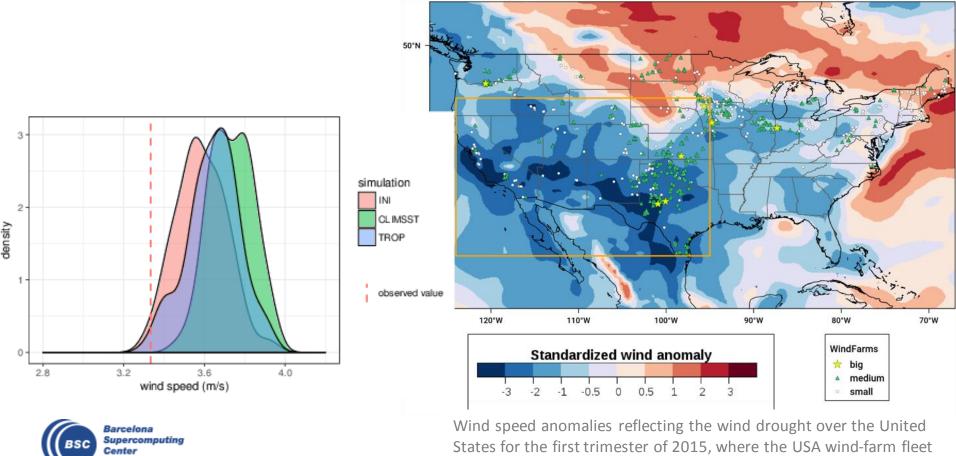
	Oct	Nov	Dec
RPSS	0.23	0.25	0.24
BS P10	-0.18	-0.23	-0.16
BS P90	0.06	0	0.03
CRPSS	0.11	0.08	0.08
EnsCorr	0.5	0.45	0.42



User engagement and forecast quality

## **Attribution** as a service

During the first quarter of 2015 the United States experienced a widespread and extended episode of low surface wind speeds. Some wind farms did not generate enough cash for their steady payments, decreasing the value of wind farm assets.



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is also shown (Lledó et al., JGR 2018) Lledó et al. (2017, JGR)

## Climate data and forecast quality Observational uncertainty in verification

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

The election of a specific dataset as a reference for the forecast verification can lead to different results.

-0'7

-0.8

-0.6 -0.5 -0.4 -0.3

-0.2

-0'1

# Ega-Interim Correlation

0.2

0.1

0.3

0.4



0.6

0.7

0.8

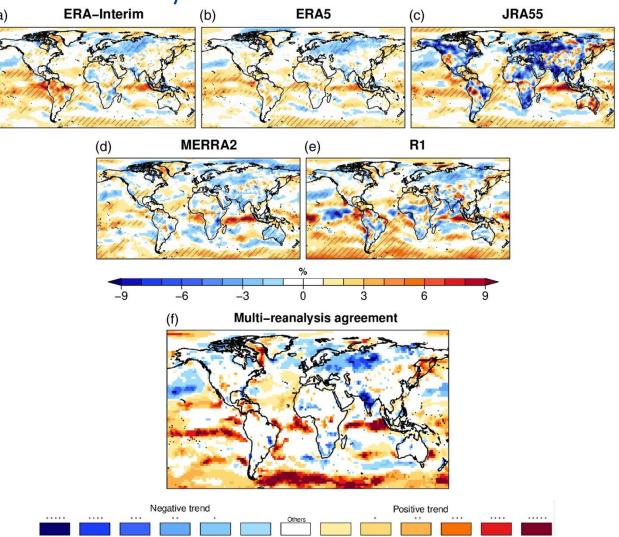
0.9

0.5

## Climate data and forecast quality

# **Observational uncertainty relevant to users**

10-m wind speed variability (in percentage of the mean wind) for the multireanalysis (MR) and five reanalyses in DJF over 1981-2017.



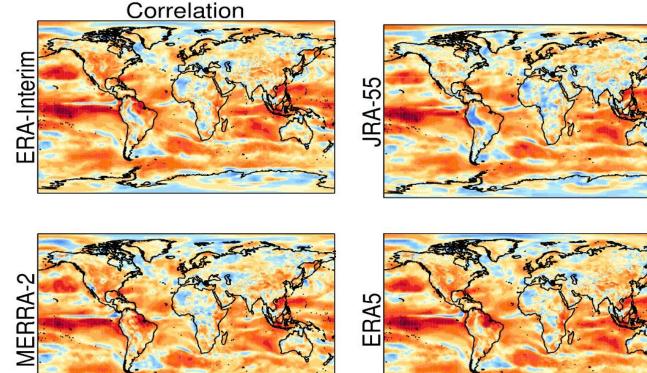


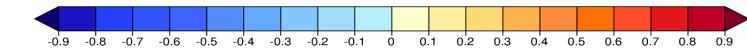
Ramon et al. (2019, QJRMS)

## Climate data and forecast quality Observational uncertainty in verification

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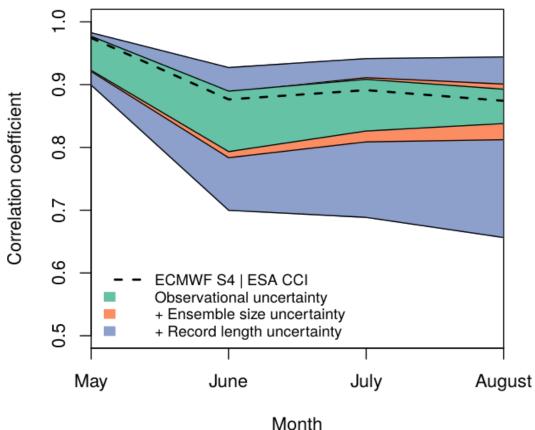




Climate data and forecast quality

# **Sources of uncertainty of forecast quality**

Niño 3.4 SST correlation of the ensemble mean for EC-Earth3.1 (T511/ORCA025) predictions with ERAInt and GLORYS2v1 ics, and BSC sea-ice reconstruction started every May over 1993-2009.



Prediction skill ENSO

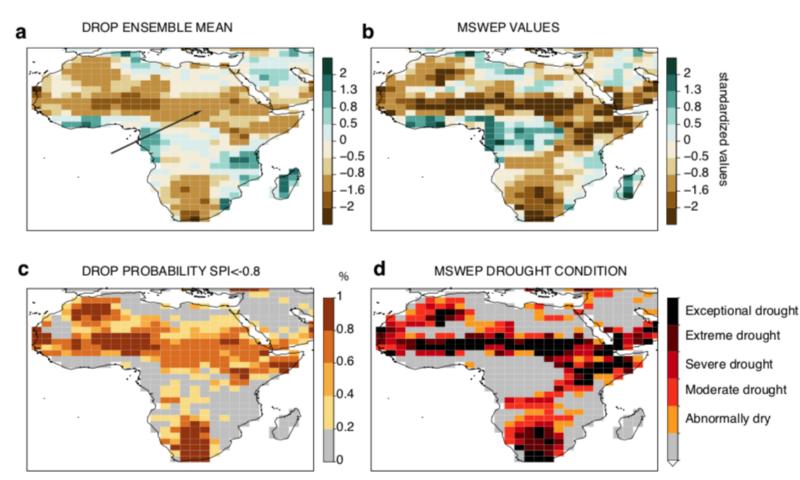


Bellprat et al. (2017, Rem. Sens. Env.)

Climate data and forecast quality

# **Observational uncertainty for monitoring**

Multisource observational estimates of drought using SPEI12 for 1984. Note the probabilistic observational estimates of drought.

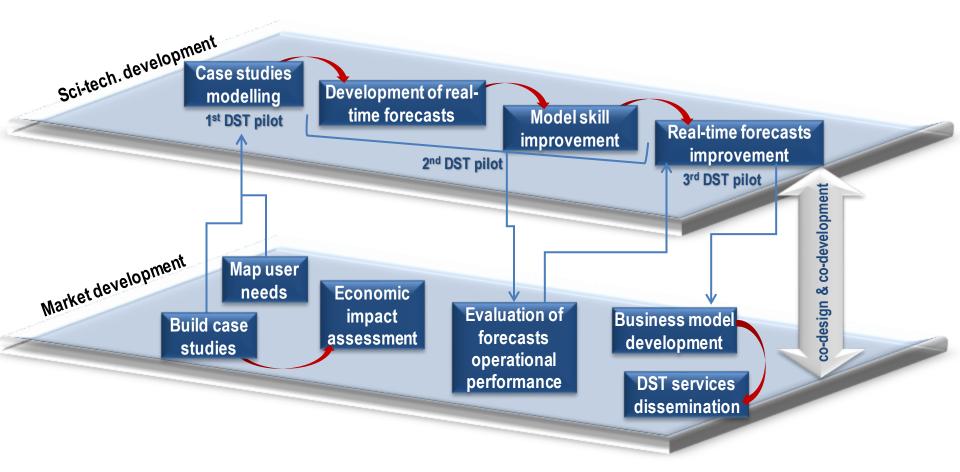




Turco et al. (2019, BAMS)

## Climate service

# The message in a transdisciplinary environment

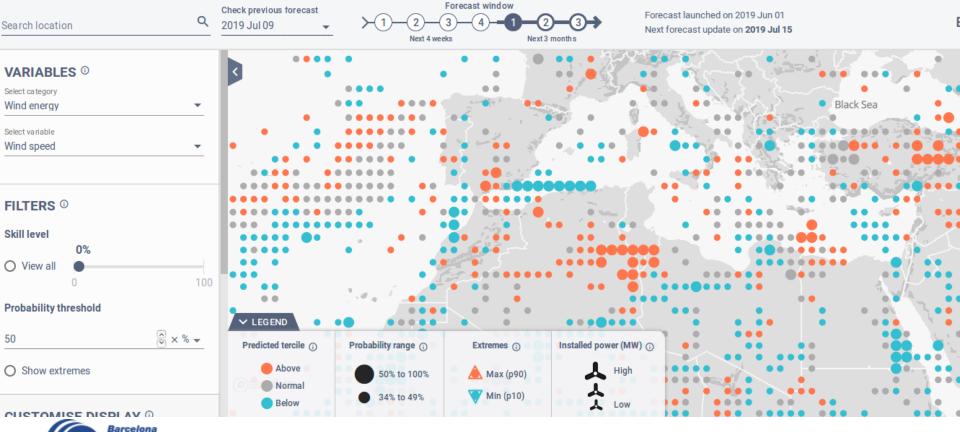




Climate service and message synthesis

# **Prototypical climate services for energy**

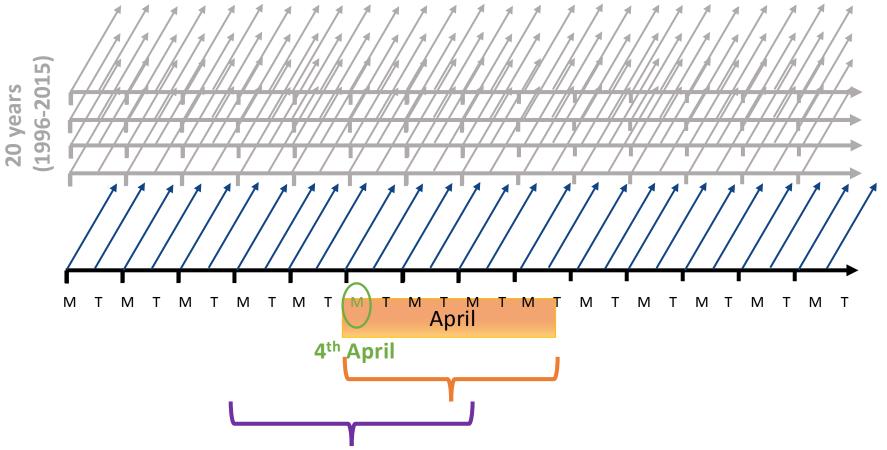
S2S4E is developing a <u>decision support tool</u> for the renewable energy sector based on <u>Copernicus climate forecasts</u> and NCEP operational predictions codesigned with the industry for periodic updates on the state of relevant climate variables.





Standards and traceability

# The (service) devil is in the detail



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





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A. Manrique

## Standards and traceability

# The (service) devil is in the detail

april Fair CRPSS - Fcst time: Days 12-18 Single startdate (raw) Single startdate (simple bias) 0.9 0.8 0.7 0.6 Monthly startdates, Monthly startdates, monthly clim (raw) monthly clim (simple bias) 0.5 0.4 0.3 0.2 0.1 Monthly startdates, Monthly startdates, 0 weekly clim (simple bias) weekly clim (raw) -0.1 bias -0.2 -0.3 -0.4 -0.5 Monthly startdates. Monthly startdates, monthly clim running window (raw) monthly clim running window (simple bias) -0.6 -0.7 -0.8 -0.9

ECMWF 2016 System Period: 1996-2015 Month: April Reference: ERA Int 2-m temperature



Monthly: good skill, but ...

Monthly running window, but weekly for the bias adjustment: lower skill , and too noisy for the adjustment



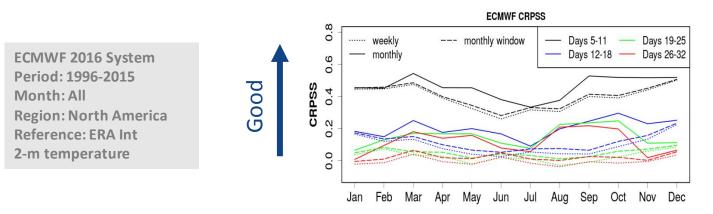
Monthly running window: more credible estimates

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## Standards and traceability

# The (service) devil is in the detail



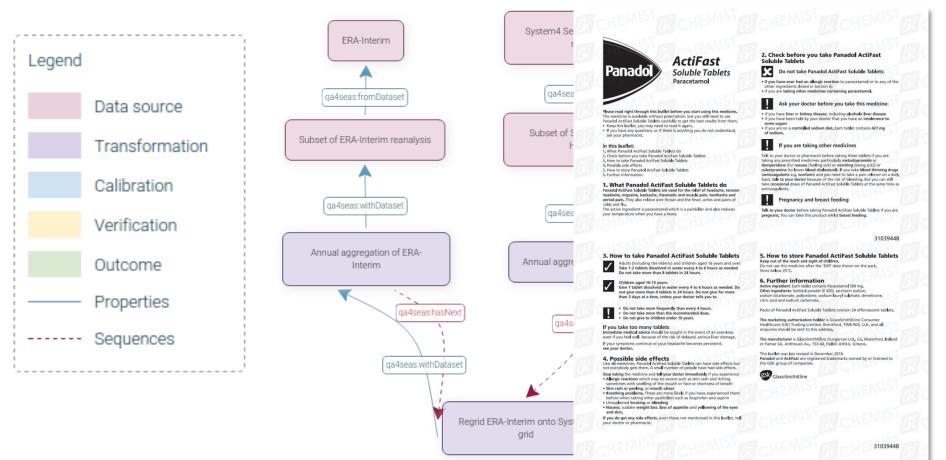


A. Manrique

#### Climate data and traceability

# How traceable are the forecast products?

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





J. Bedia (Predictia)

## Climate data and traceability

# **Evaluation and quality control**

C3S is developing the evaluation and quality control (EQC) function of the climate data store to:

- Provide a user-led overarching EQC service for the whole CDS
- Provide an independent quality assessment





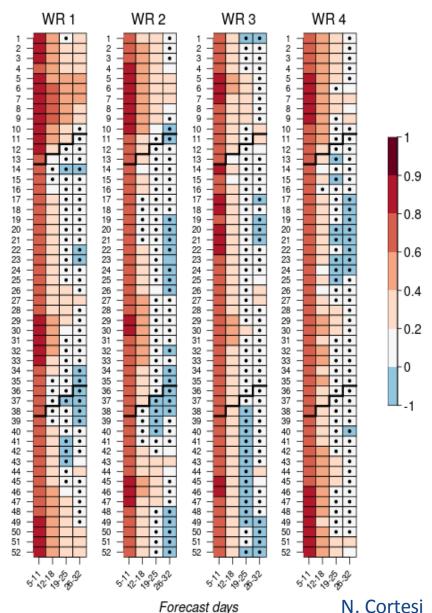


## User engagement

# A user view of the WRs

ECMWF 2018 System Period: 1998-2017 Month: All Region: North Atlantic Reference: ERA Int MSLP

Correlation between observed and predicted regime frequencies as a function of the forecast week (x-axis) and for t date each of the 52 weeks of the year (y-axis). Black ශී points for non-significant correlations using a paired t-test 0.01 at level.





User engagement

# A user view of the WRs

The influence of each regime on the wind speed conditions can be explored in terms of composite maps.

How to quantify the combined effect of the four weather regimes?

## RECONSTRUCTION

Total number of regimes (R = 4 in this example)

 $varRecon_{mon,yr}(lat, lon)$ 

PP-S

3.4

27.6%

28.3%

24%

20.2%

-8

-16

0

hPa

8

16

Variable to be reconstructed (wind speed or temperature) in a particular year (yr) and month (mon) Composite map of the variable for a specific regime (r) in the same month and in the full period (crossvalidation)

 $CM_{r,mon}(lat, lon)$ 

freq<sub>r,mon,yr</sub>

Percentage of days that particular month, year and regime

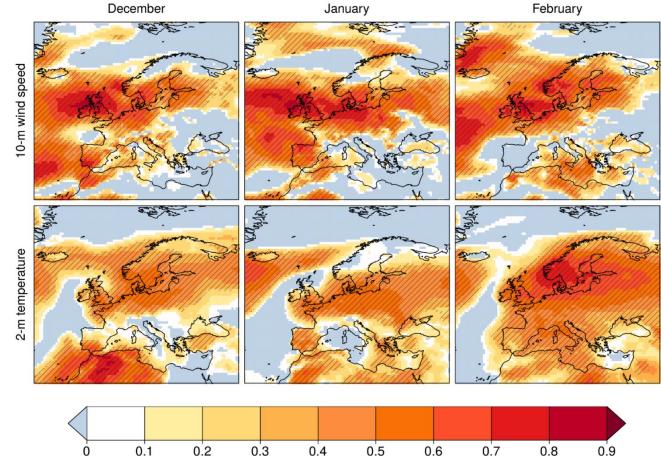
V. Torralba

User engagement

# A user view of the WRs

Period: 1982-2016 Hatching: significant at 95% level (t-test)

# Correlation coefficient between the variables reconstructed with WRs and the original ones in ERA-Interim



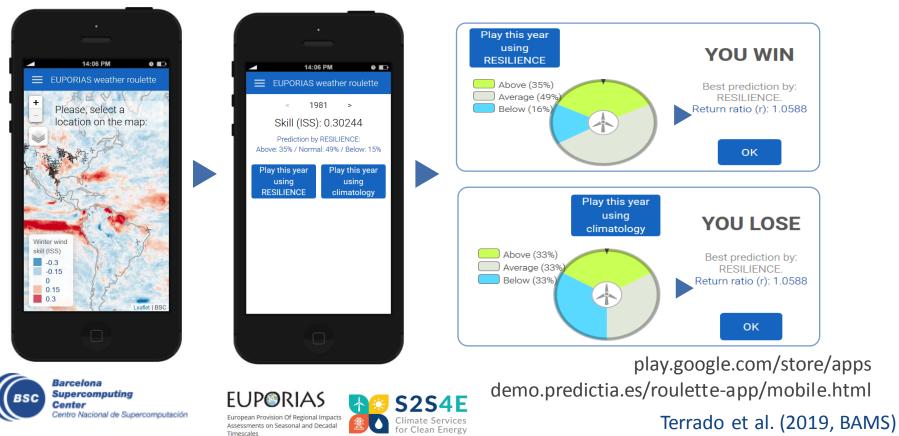
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# The communication challenge

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

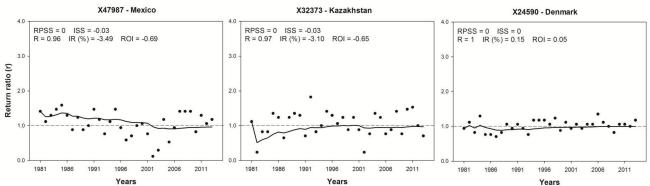
- Play against a reference taken from climatological frequencies.
- The bets are proportional to the predicted probabilities.
- The amount invested in the observed category is multiplied by 3.

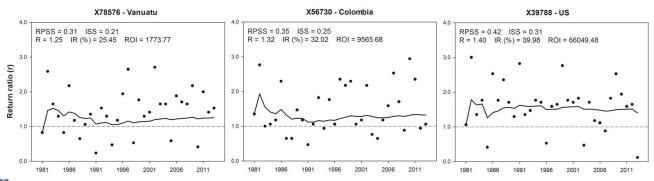


# **Illustrating prediction value**

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).



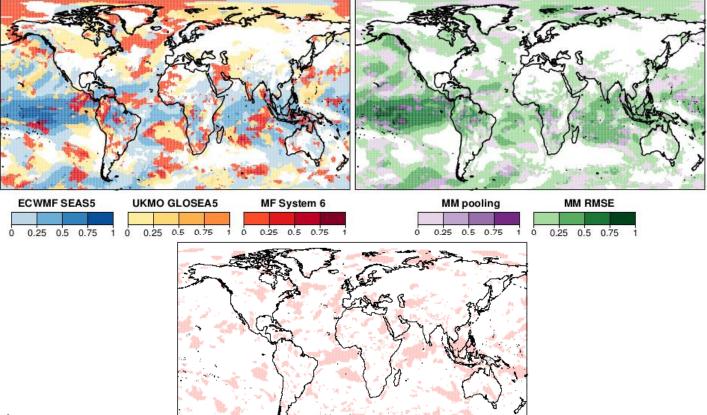




Synthesis

# **Multi-model forecasts**

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.





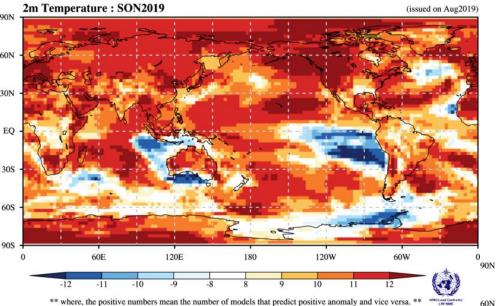
Best MM - Best indiv

V. Torralba

# **Other approaches for the synthesis**

#### **Consistency Map**

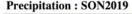
CPTEC, ECMWF, Exeter, Melbourne, Montreal, Moscow, Offenbach, Pretoria, Seoul, Tokyo, Toulouse, Washington



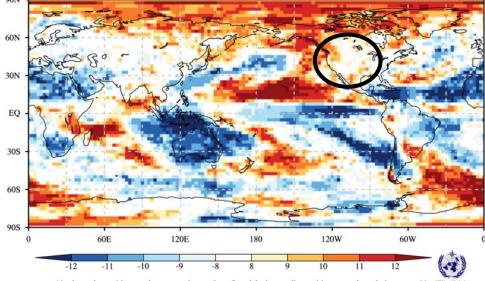
Storylines built on process-based narratives are a useful tool to communicate uncertainty when the signal is weak, the forecast is not Gaussian or windows of opportunity appear.

#### **Consistency Map**

CPTEC, ECMWF, Exeter, Melbourne, Montreal, Moscow, Offenbach, Pretoria, Seoul, Tokyo, Toulouse, Washington



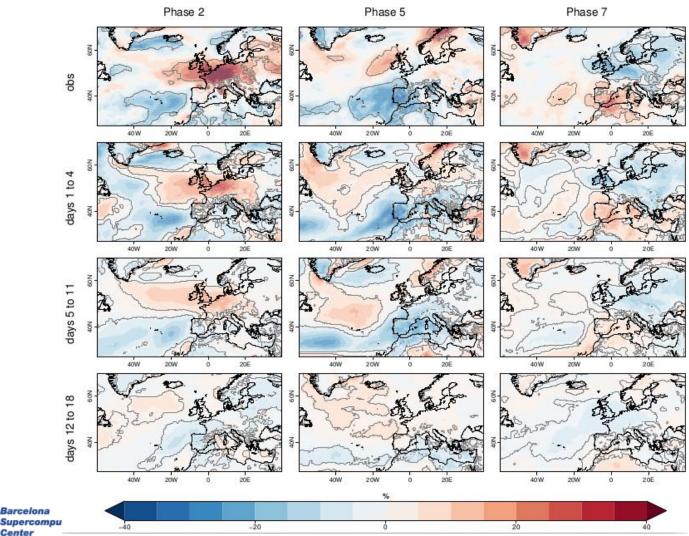
(issued on Aug2019)





# **Other approaches for the synthesis**

10-metre wind speed anomalies (percentage wrt to the mean) associated with three phases of the MJO (S2S definition) in ERA Interim and ECMWF 2016.





Ll. Lledó

# Summary

- Requests for contextual climate information based on climate forecasts come from a broadening range of users and needs to be addressed from an operational climate services perspective. Addressing this requirement require a new paradigm for climate research.
- Applications struggle with the current compartmental provision of climate data and information.
- Entry-level documentation and training, formulation of standards, as well as communication, among many others, have become fundamental.
- None of this will materialise without appropriate investment in observational networks, increased collaboration and research for the reduction of all aspects of model error, among many other critical aspects.



