



S2S4E

Climate Services
for Clean Energy

Climate services for clean energy: the S2S4E project

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Context and motivation

Both energy supply and demand are strongly influenced by weather conditions and their evolution over time in terms of climate variability and climate change.

Like 15M

Thursday, Aug 30th 2018 1PM 25°C 4PM 26°C 5-Day Forecast

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Britain's turbines are producing 40% less energy as wind 'disappears' for six weeks across the UK causing record low electricity production

- Britain got 15 per cent of its power from wind last year — twice as much as coal
- Since the start of June, wind farms have been producing almost no electricity
- The 'wind drought' has seen July 2018 be 40% less productive than July 2017
- In the still weather, solar energy has increased by 10% to help cover the drop-off



By [JOE PINKSTONE FOR MAILONLINE](#)

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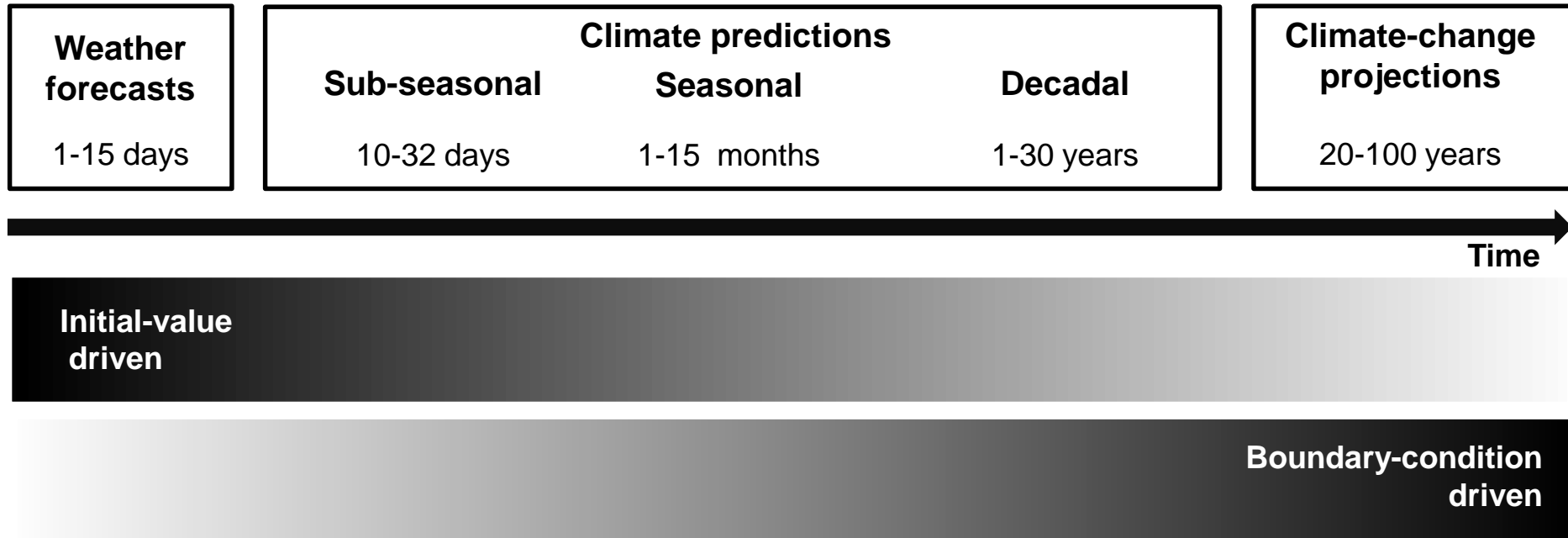
Context and motivation

▶ Energy sector routinely uses weather forecast up to several days. Beyond this time horizon, climatological data are used.



Met mast on Gwynt y Môr offshore wind farm (source: solar wheel)

Climate predictions



Adapted from: Meehl et al. (2009)

Predictability

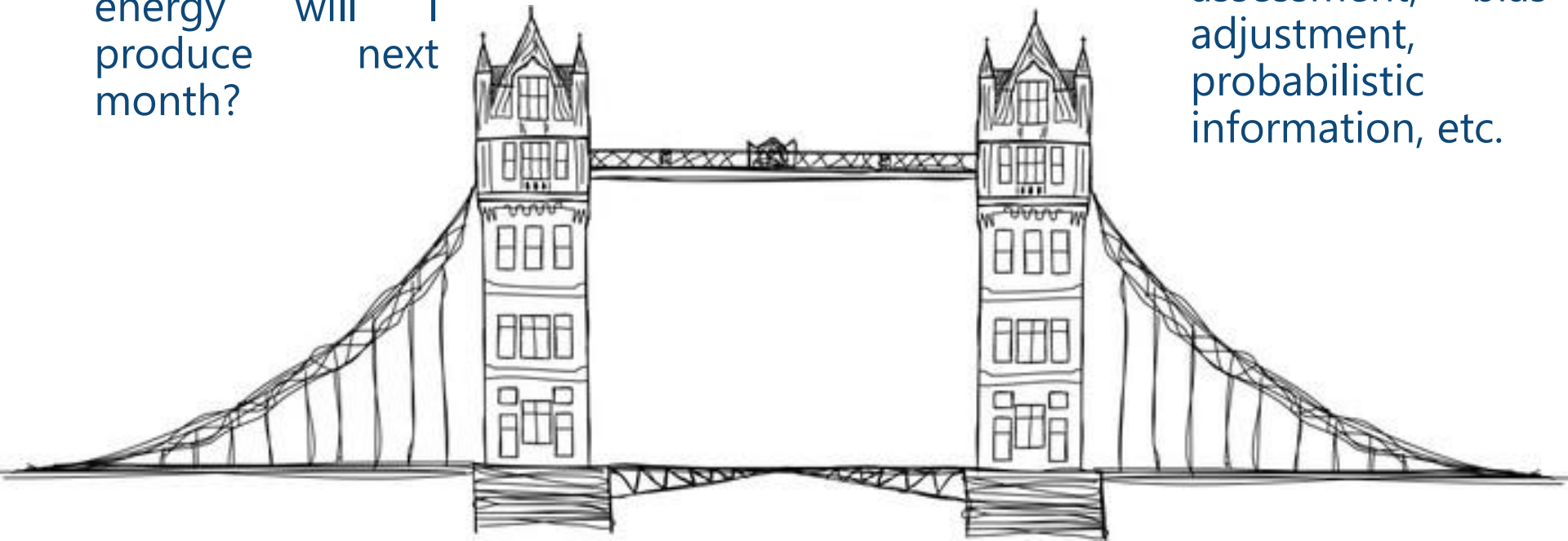
▶ How can we predict climate for the coming season if we cannot predict the weather next week? Slow components (sea surface temperature, soil moisture, etc.) force the atmosphere.



Climate services

▶ User: How much energy will I produce next month?

▶ Scientist: Skill assessment, bias adjustment, probabilistic information, etc.

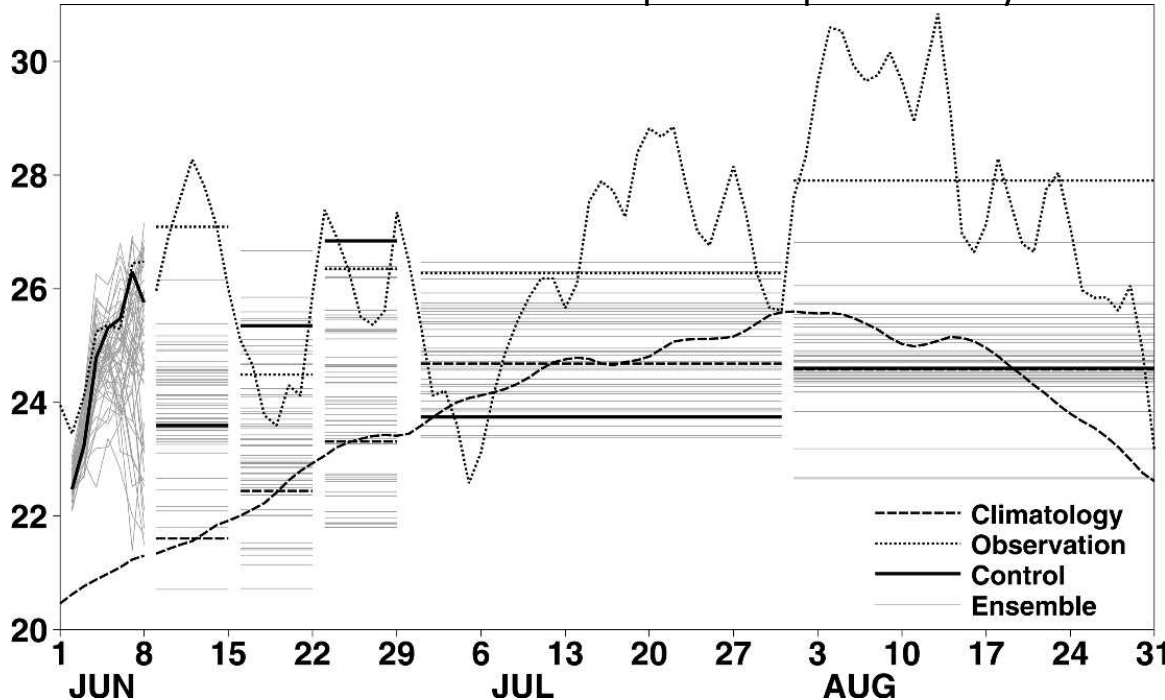


S2S4E objective

Objective

► S2S4E will offer an innovative service to improve RE variability management by developing new research methods exploring the frontiers of weather conditions for future weeks and months. The main output of S2S4E will be a user co-designed Decision Support Tool (DST) that for the first time integrates sub-seasonal to seasonal (S2S) climate predictions with RE production and electricity demand.

Heat wave 2003. Prediction of temperature produced by ECMWF



Observations (dotted) and forecasts (solid) made by ECMWF at the beginning of June of European 2-m land temperatures (C).

Source: Rodwell and Doblas-Reyes, 2006

Applications



Applications for wind/solar/hydro generation

Post-construction decisions
Energy producers: commit energy sales for next day
Grid operators: Market prices and grid balance
Energy traders: Anticipate energy prices
Plant operators: planning for cleaning and maintenance

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 investments

Pre-construction decisions
Power plant developers: Site selection. Future risks assessment.
Investors: Evaluate return on investments
Policy-makers: Asses changes to energy mix
River-basin managers: understand changes to better manage the river flow



Applications for demand
Daily operation decisions
Grid operators: Anticipate hot/cold days. Schedule power plants to reinforce supply.
Energy traders: Anticipate energy prices.

Mid-term planning
Grid operators: Anticipate hotter/colder seasons. Schedule power plants to reinforce supply.
Energy traders: Anticipate energy prices.

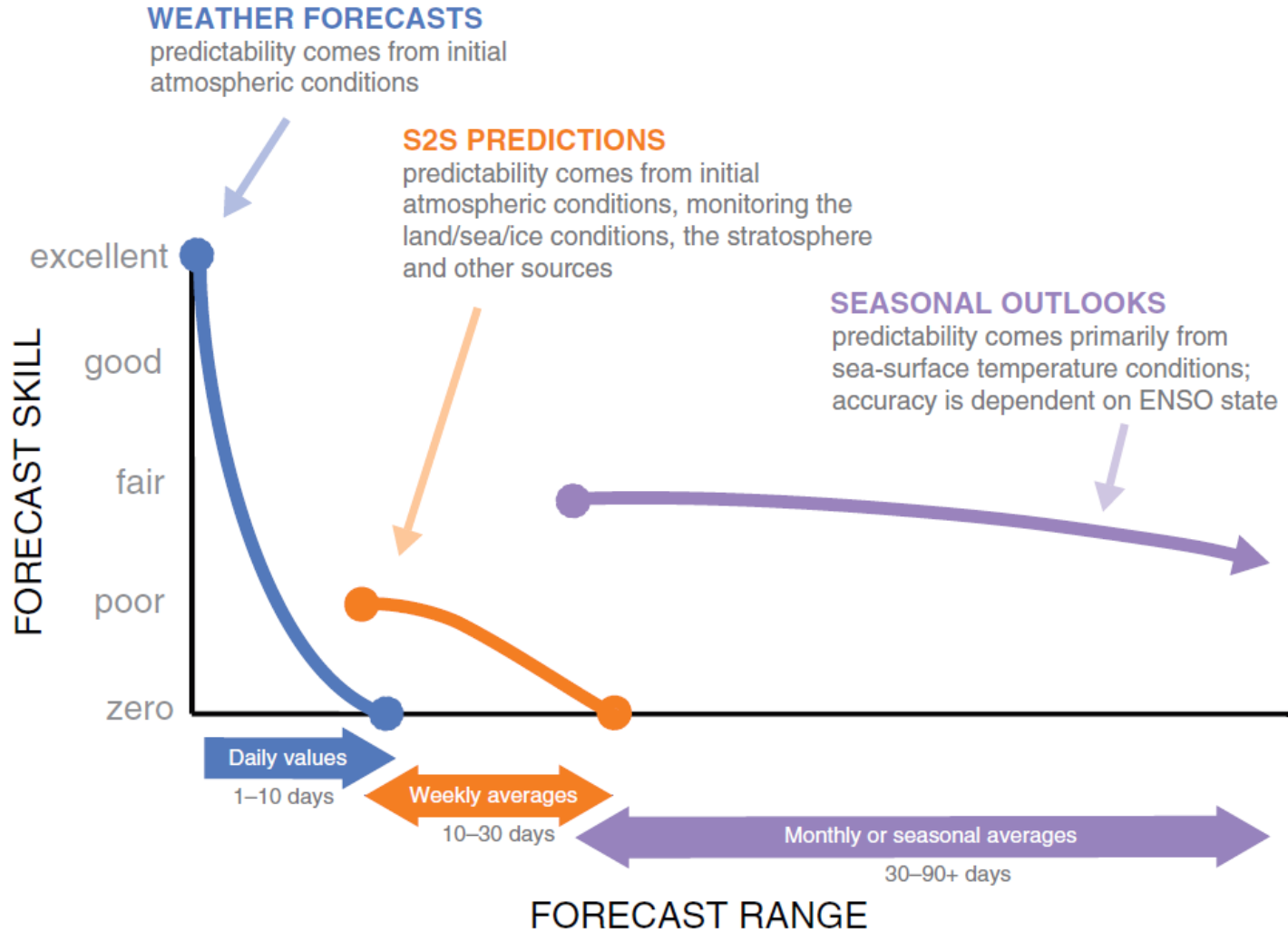
Long-term planning
Grid operators: Anticipate addition of more capacity. Adaptation of transmission lines
Policy-makers: Plan addition of more capacity. Understand changes to energy mix



S2S4E project

Challenges and opportunities

S2S Forecast range and skill

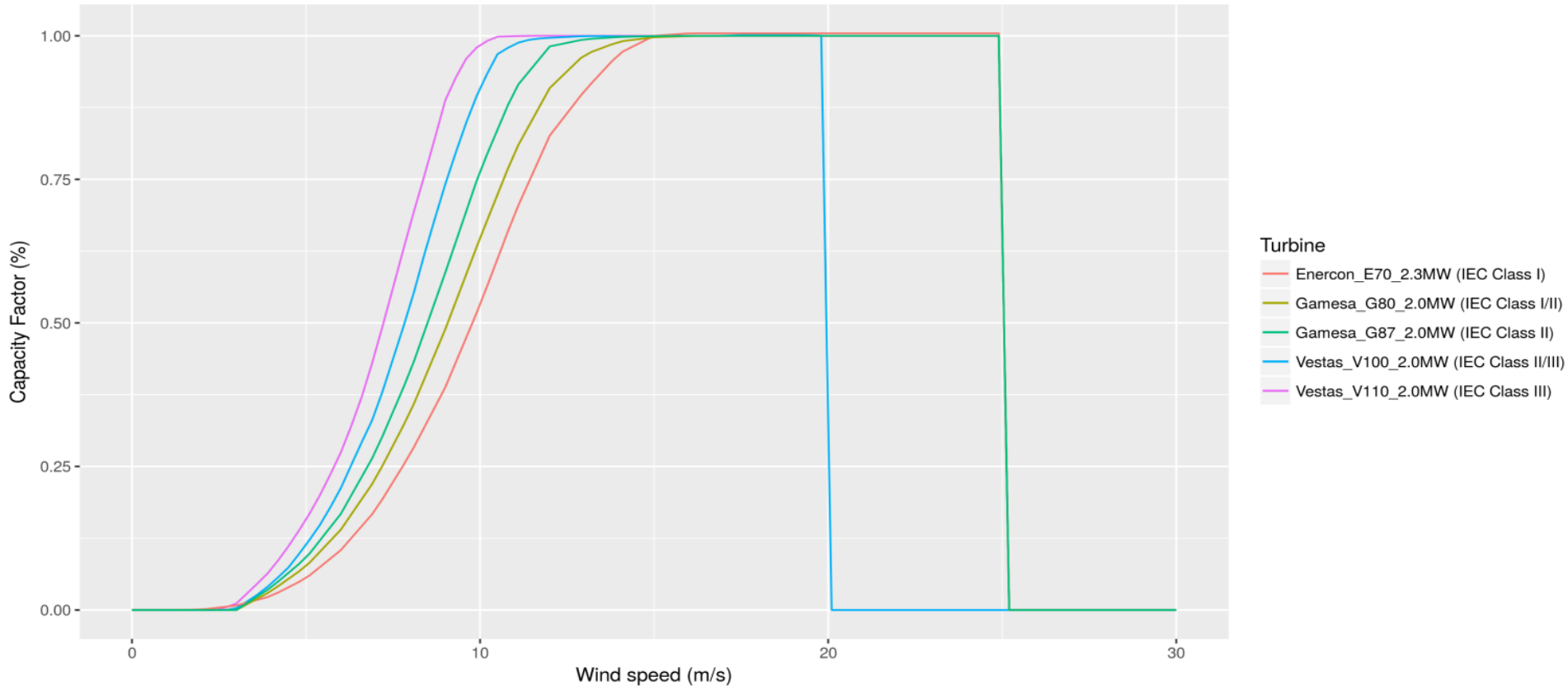


Qualitative estimate of forecast skill based on forecast range from short-range weather forecasts to long-range seasonal predictions, including potential sources of predictability. Relative skill is based on differing forecast averaging periods. (Source: White et al., 2017)

Co-development

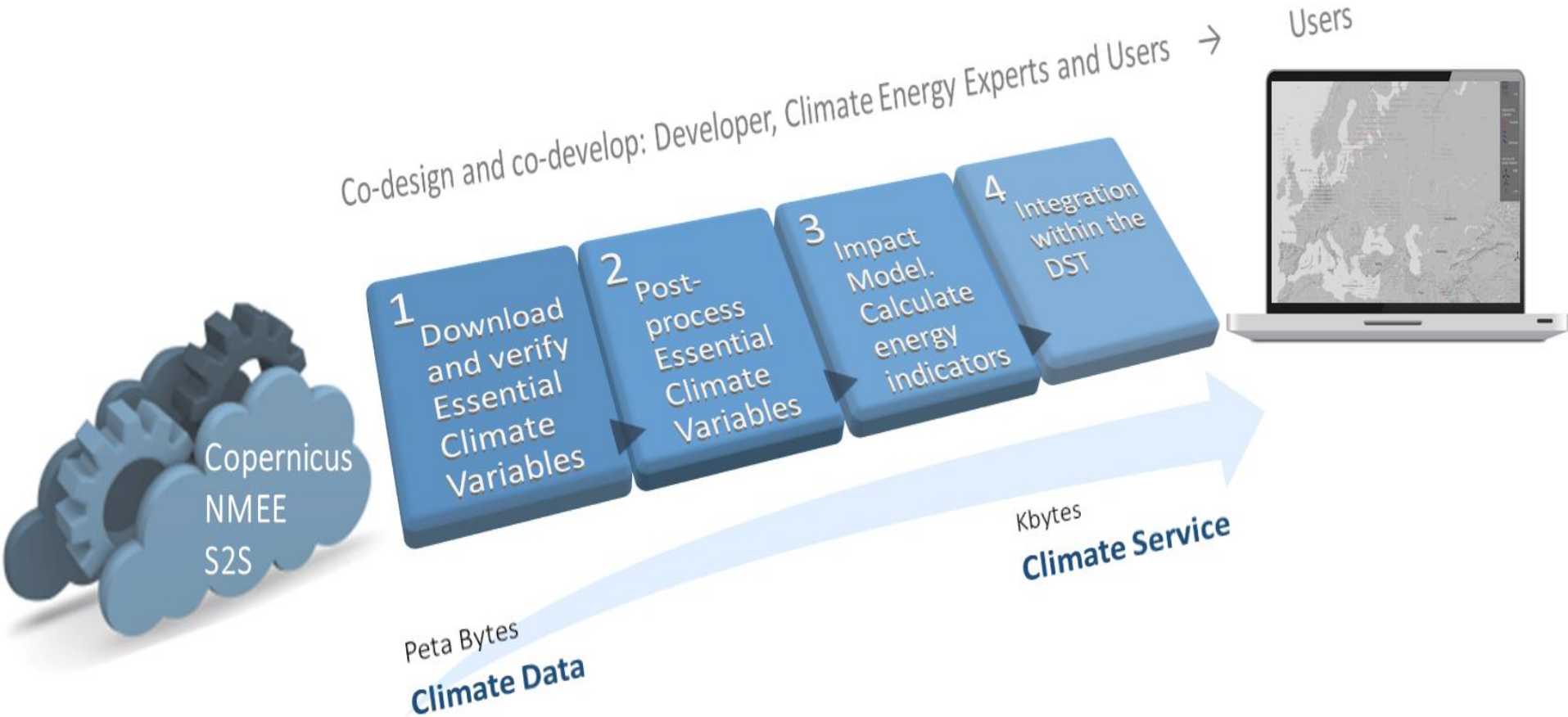
▶ E.g.: development of useful and tailored indicators

Selected power curves



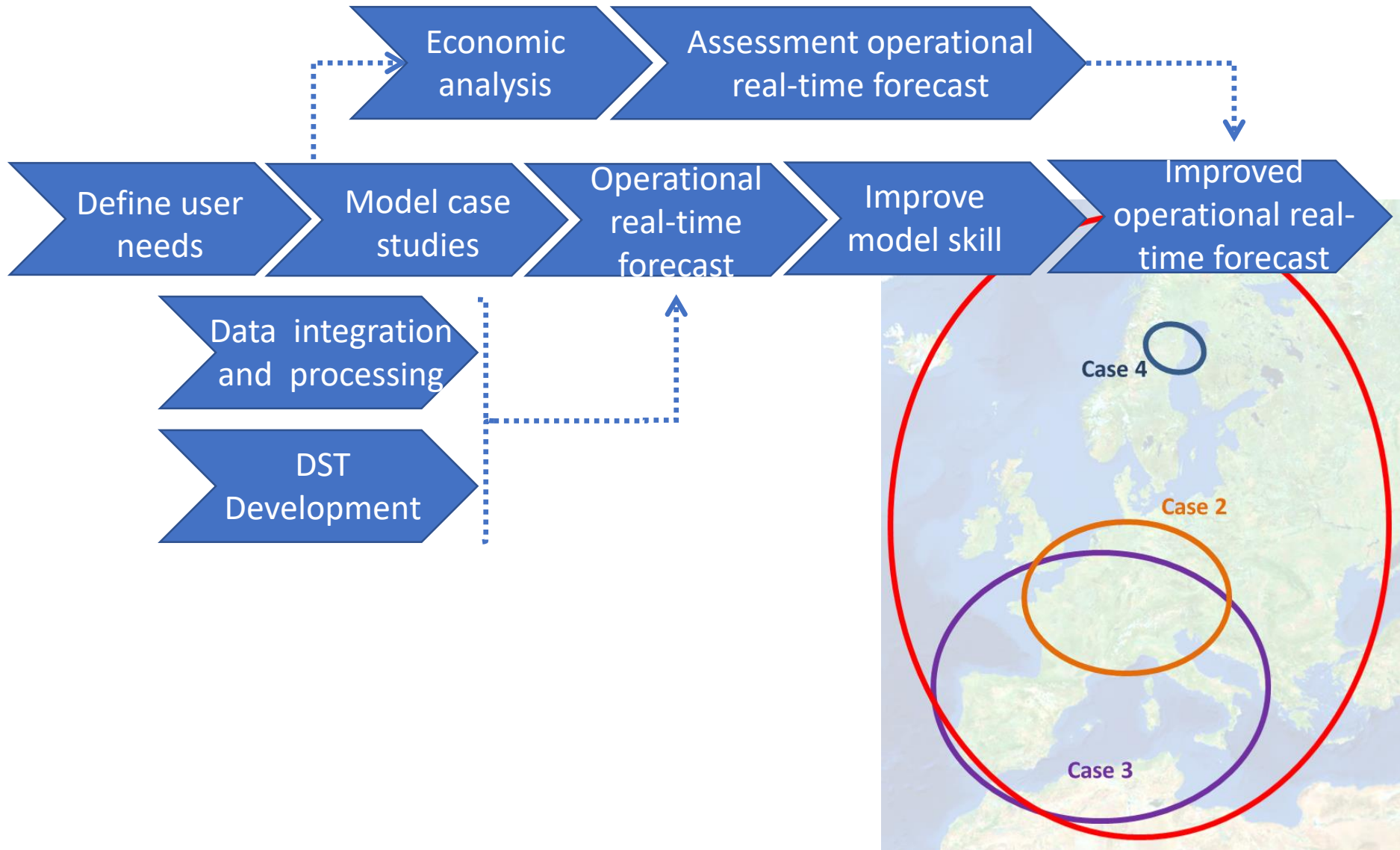
Power curves for three turbines representing IEC classes.

From data to service



Methodology and firts results

Methodology

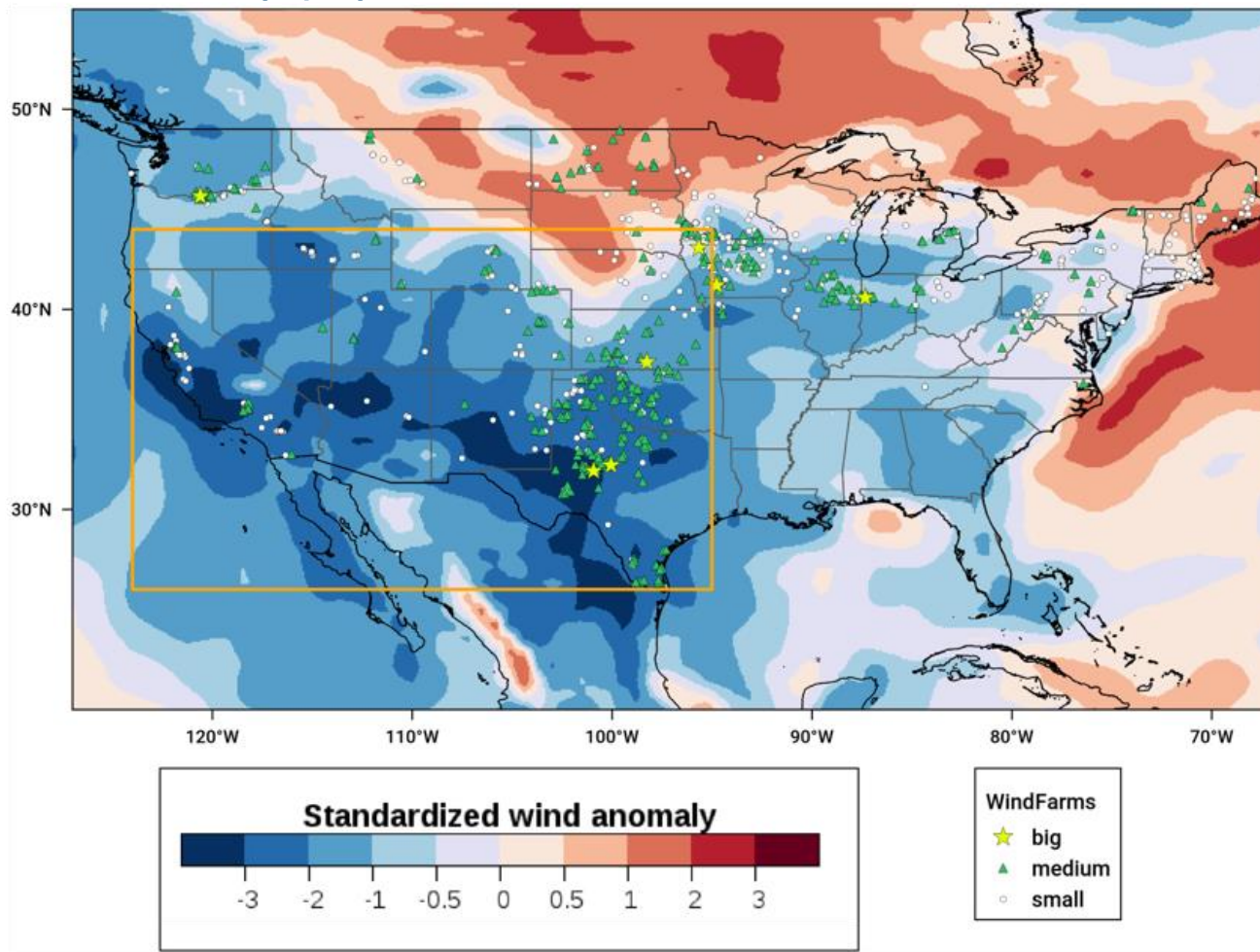


Case study: wind drought in US

Lledó et al., 2018: Investigating the effects of Pacific sea surface temperatures on the wind drought of 2015 over the United States. *Journal of Geophysical Research*

Wind drought in US

During the first quarter of 2015 the United States experienced a widespread and extended episode of low surface wind speeds. This episode had a strong impact on wind power generation. Some wind farms did not generate enough cash for their steady payments, and the value of wind farm assets decreased.

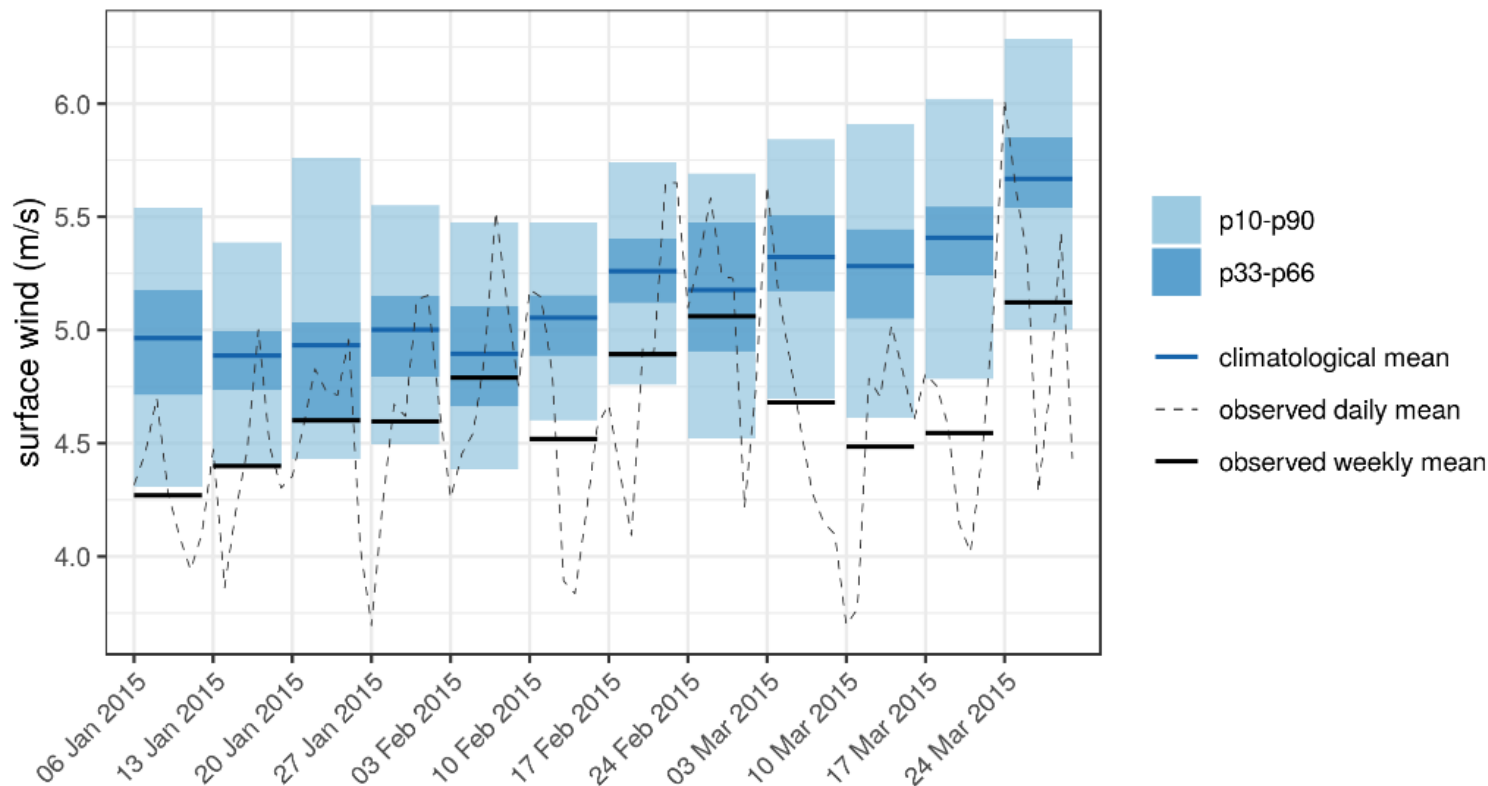


Wind speed anomalies reflecting the wind drought over the United States for the first trimester of 2015. The US wind farm fleet is also shown.

Wind drought in US

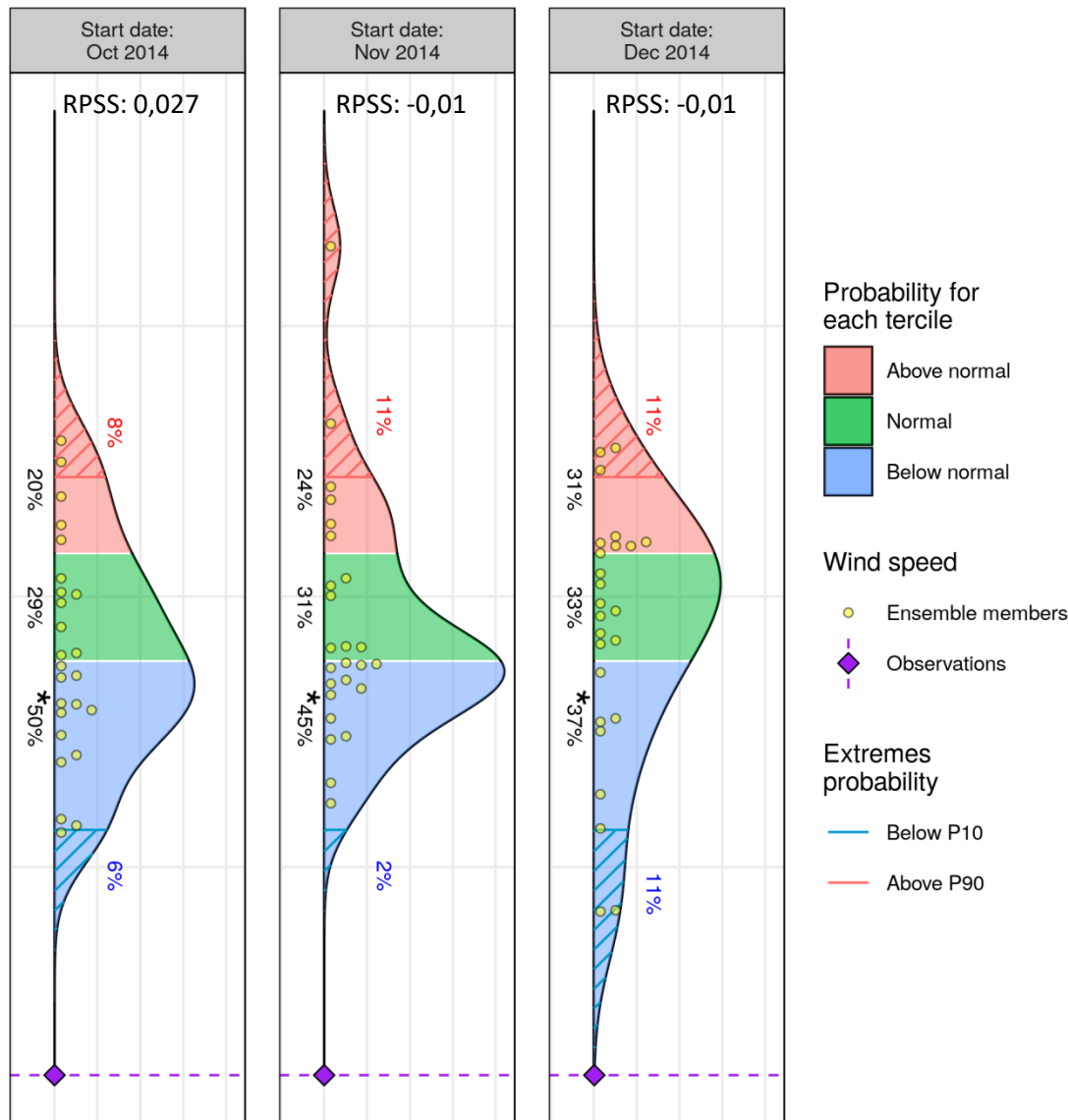
During the first quarter of 2015 the United States experienced a widespread and extended episode of low surface wind speeds. This episode had a strong impact on wind power generation. Some wind farms did not generate enough cash for their steady payments, and the value of wind farm assets decreased.

Observed weekly means and climatology



Available seasonal forecast

Forecasts for Jan-Mar 2015



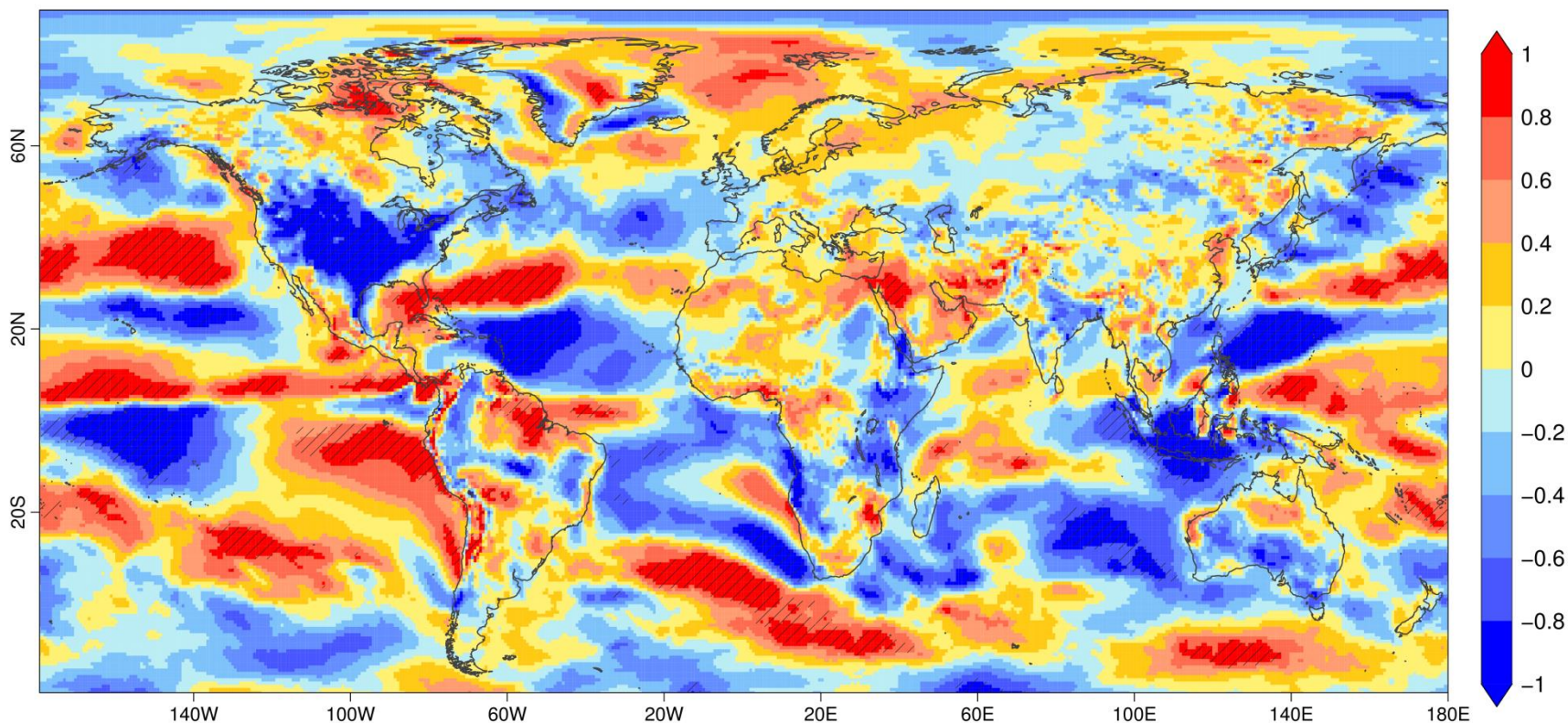
System: ECMWF S4
Reanalysis: ERA-Interim
Bias adjusted –calibrated
Hindcast: 1993-2015



Which decisions would you take in view of those forecasts?

NIÑO3.4 teleconnection

ERA-Interim / 10m wind speed / NIÑO3.4 positive minus neutral impact
DJF / 1981-2015

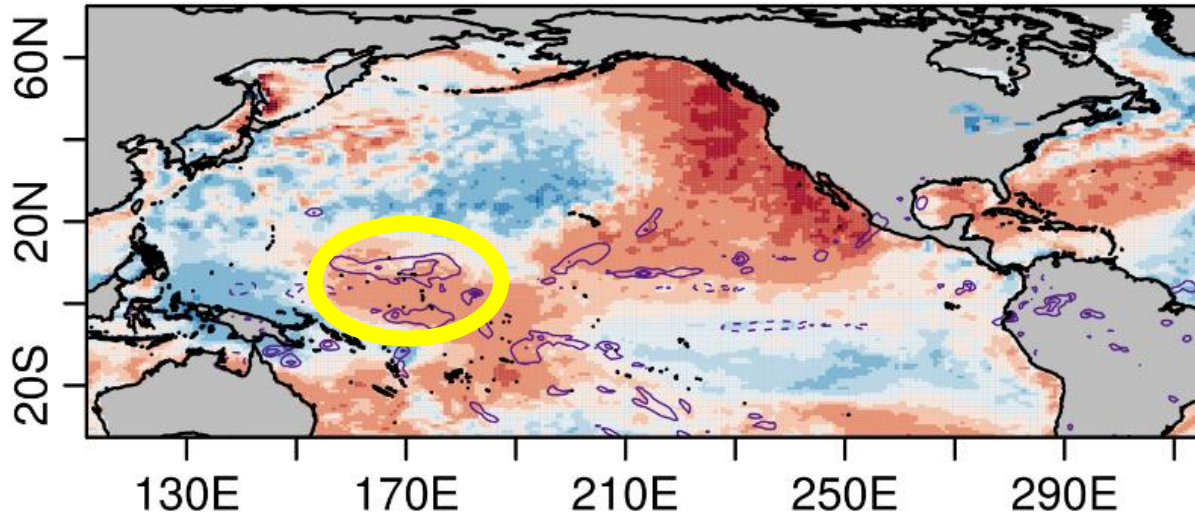


Bias correction: none
Hatched area: significant at 95% confidence level from a two tailed Student's t-test
Mask: sea depth below 50m

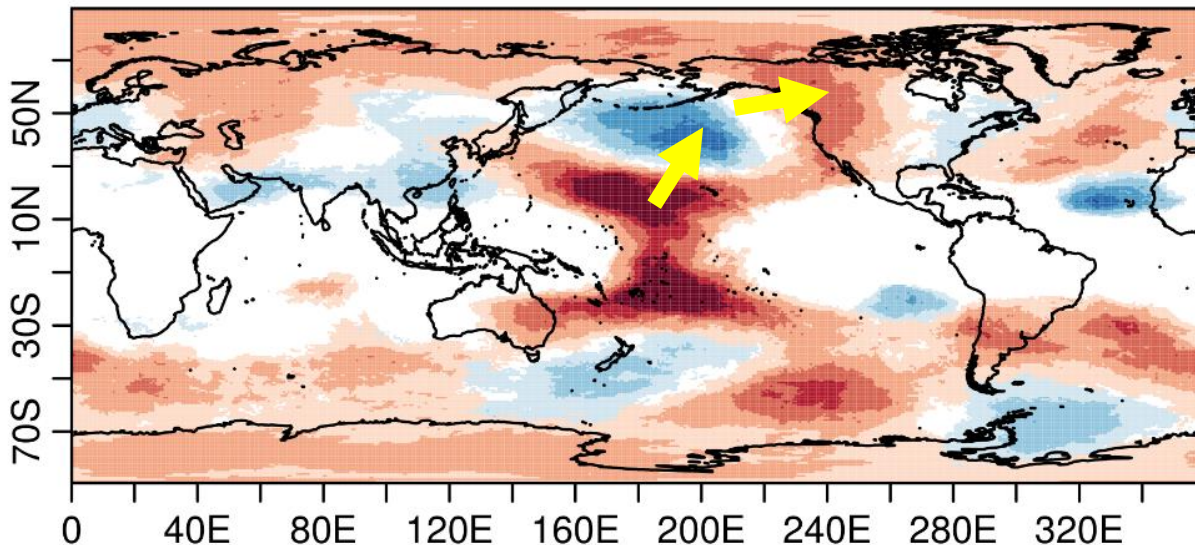
Impact maps between NIÑO3.4 teleconnection index 10m wind speed from ERA-Interim reanalysis.

Causes

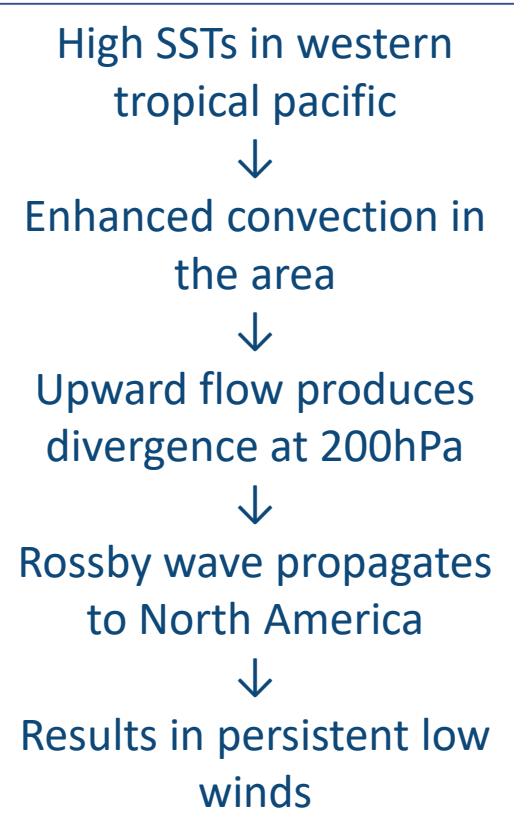
January-March 2015



SST and precip



GH @200hPa

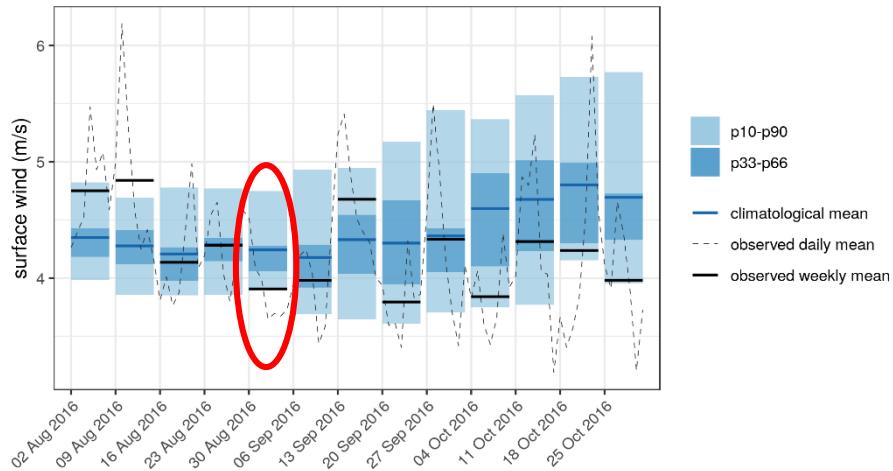


Case study: heat wave and wind drought in Spain. Sep 2016

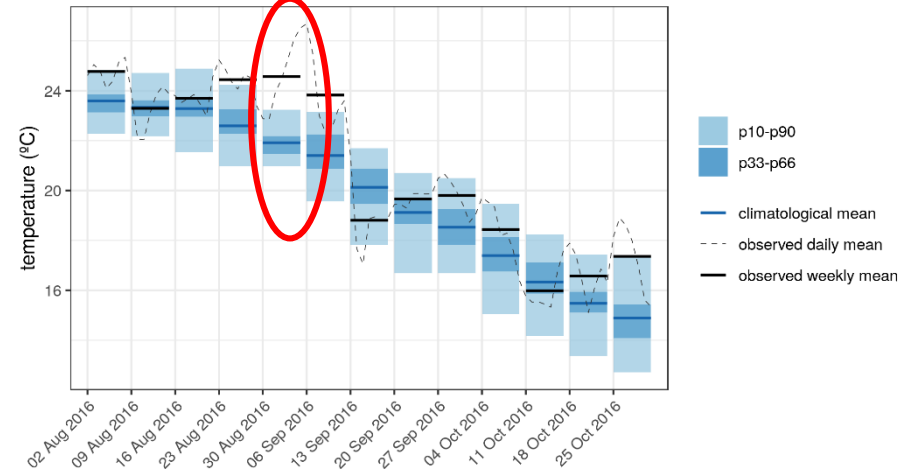
Heat wave and wind drought in Spain. Sep 2016

The cold spell over Europe created a combination of large increase in electricity demand and lower than usual hydro and wind power generation.

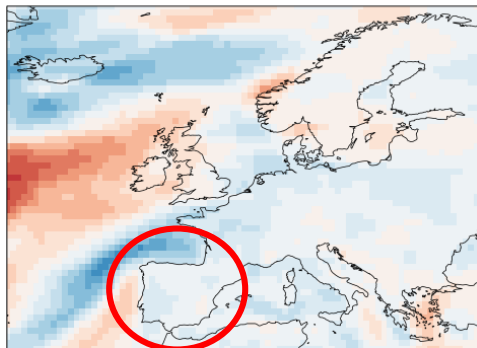
Observed weekly means and climatology



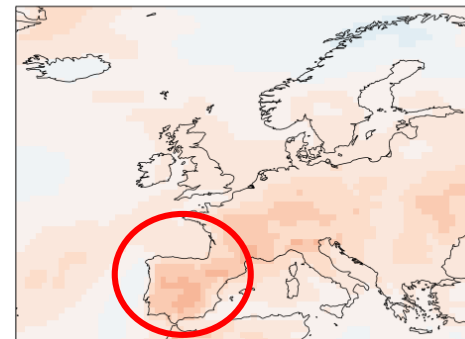
Observed weekly means and climatology



2016-08-30



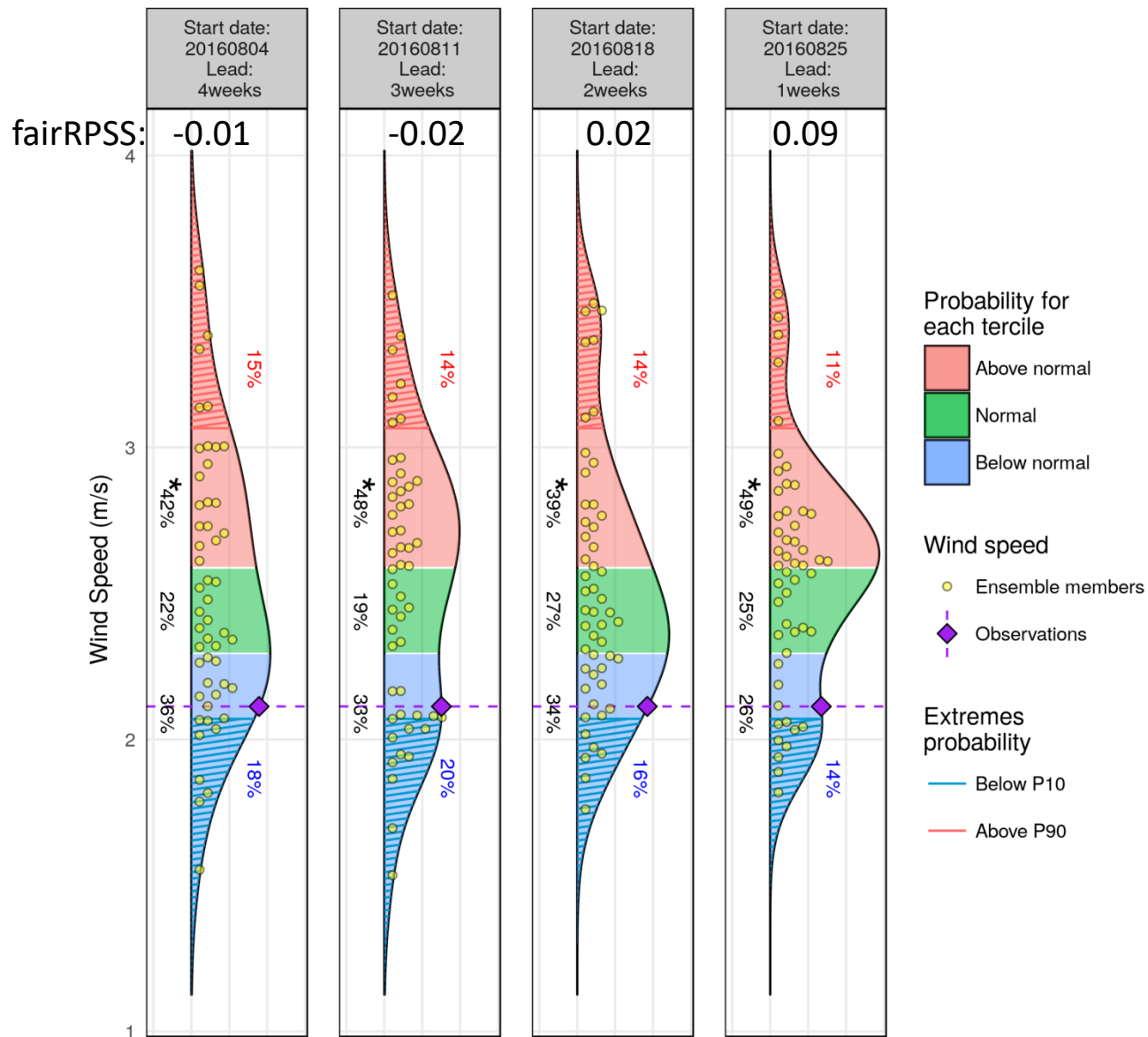
2016-08-30



Surface wind and temperature anomalies. With respect climatology (1981-2017)

Forecast available: wind speed

Forecasts for week starting 2016-08-30

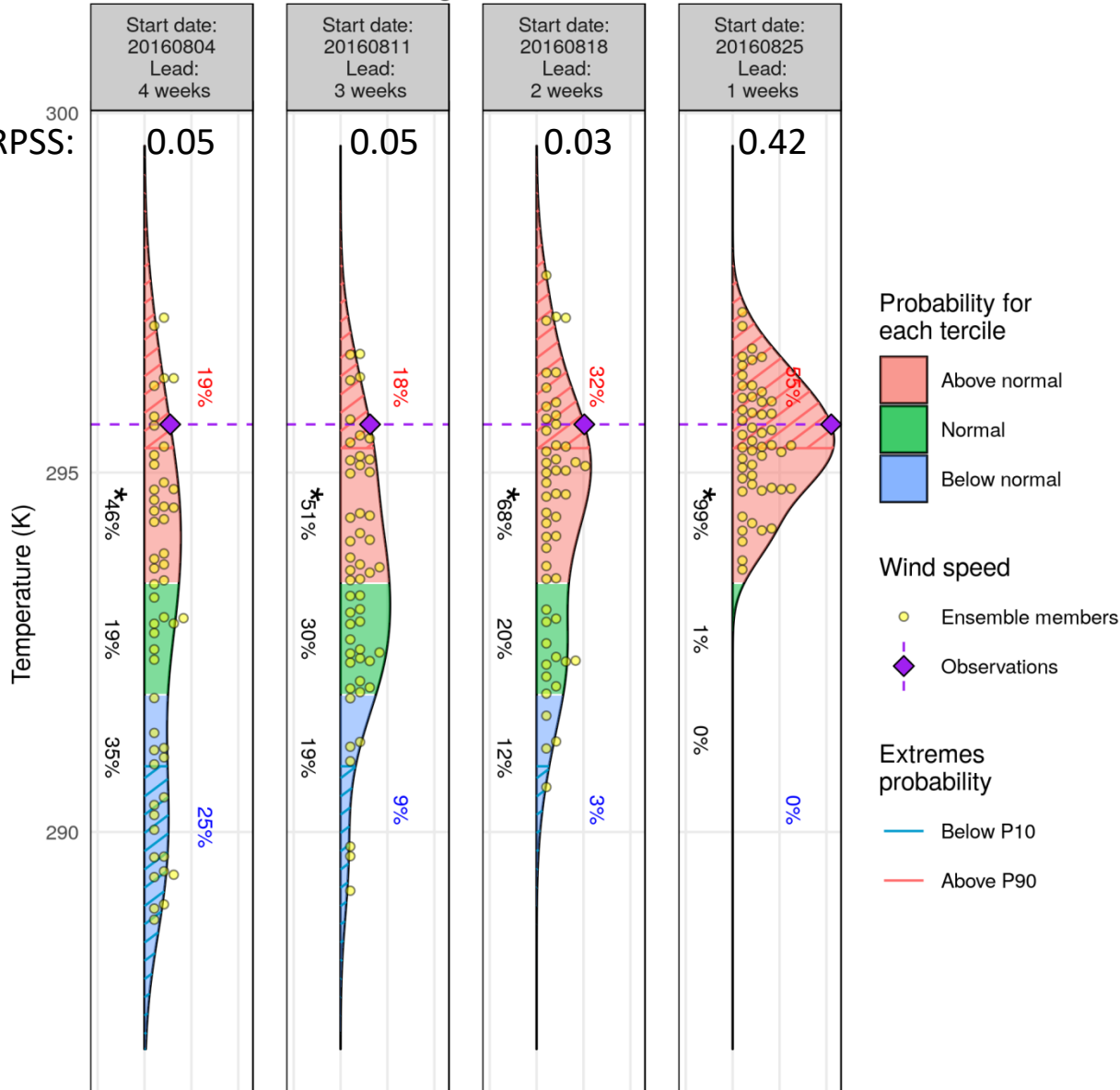


System: ECMWF monthly prediction system
 Reanalysis: ERA-Interim
 Bias adjusted –calibrated
 Hindcast: 1996-2015
 Lat= 40.5 N/Lon = 358.5 E

Forecast available: temperature

Forecasts for week starting 2016-08-30

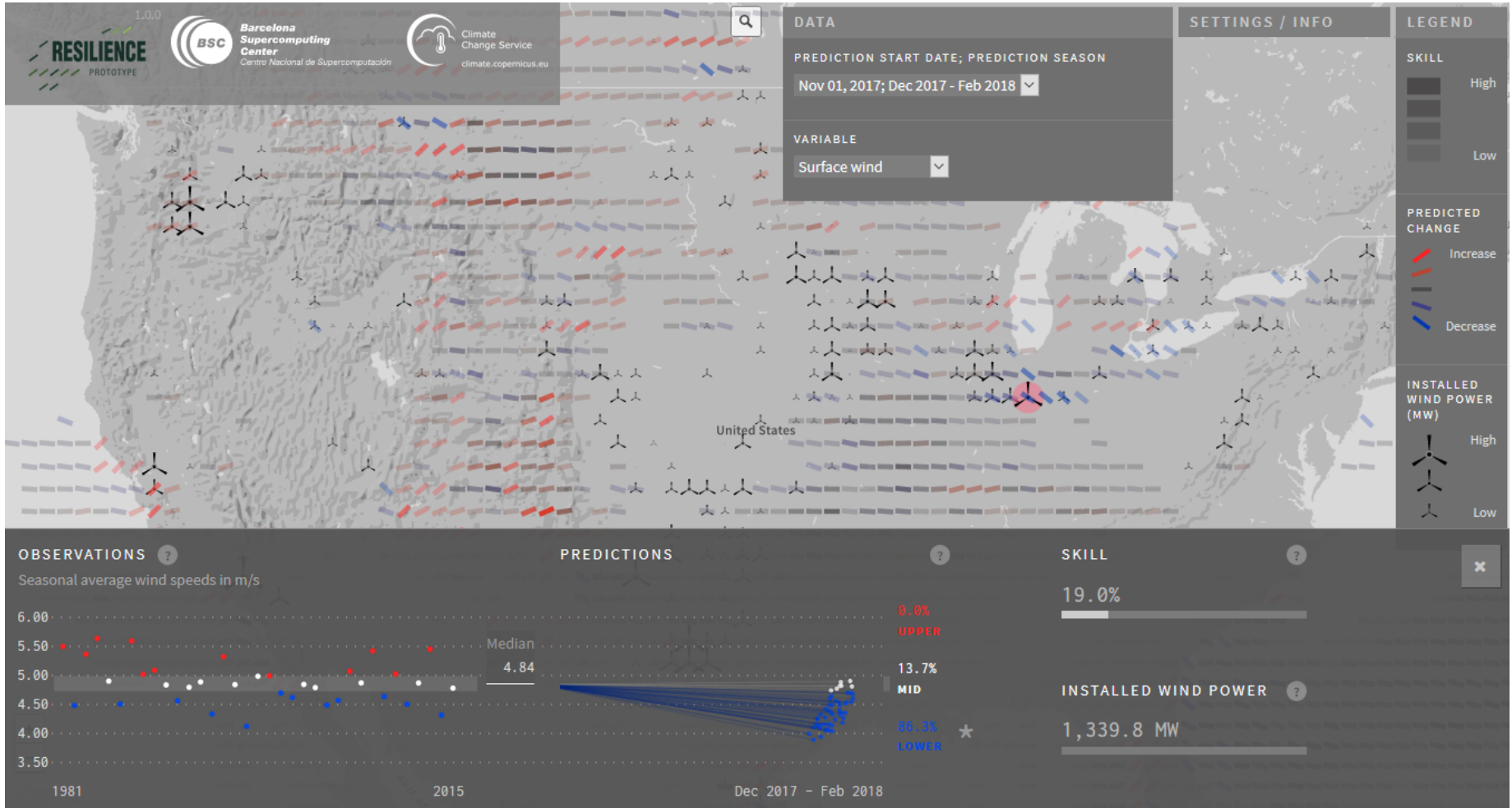
fairRPSS:



System: ECMWF monthly prediction system
 Reanalysis: ERA-Interim
 Bias adjusted –calibrated
 Hindcast: 1996-2015
 Lat= 40.5 N/Lon = 358.5 E

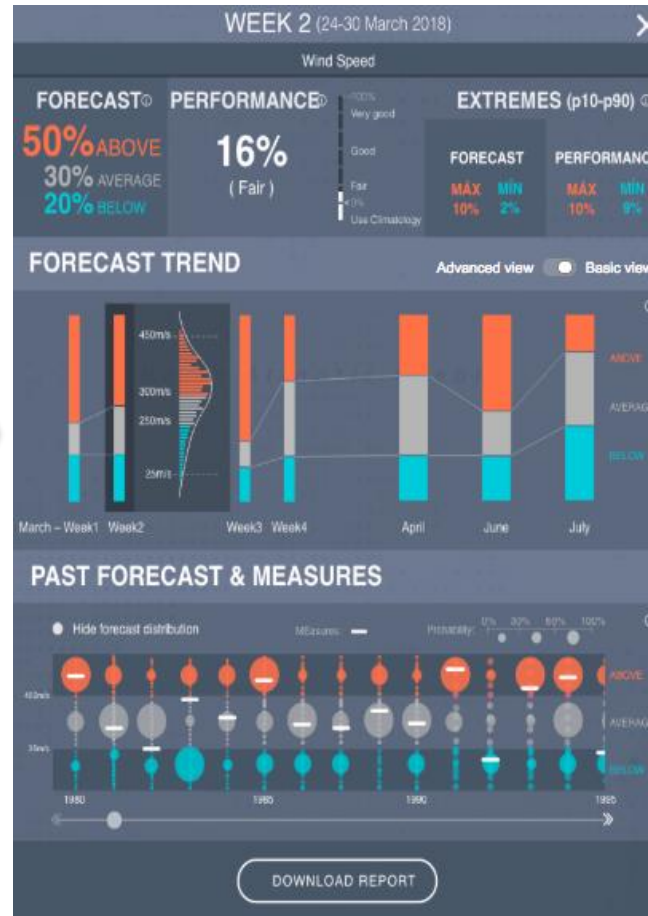
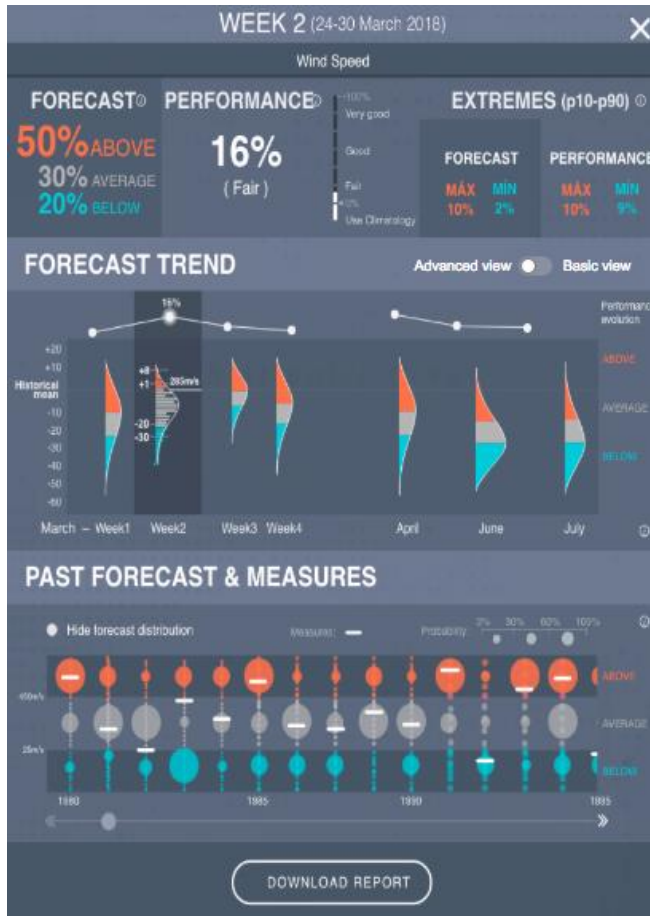
Co-development of the decision support tool

DST

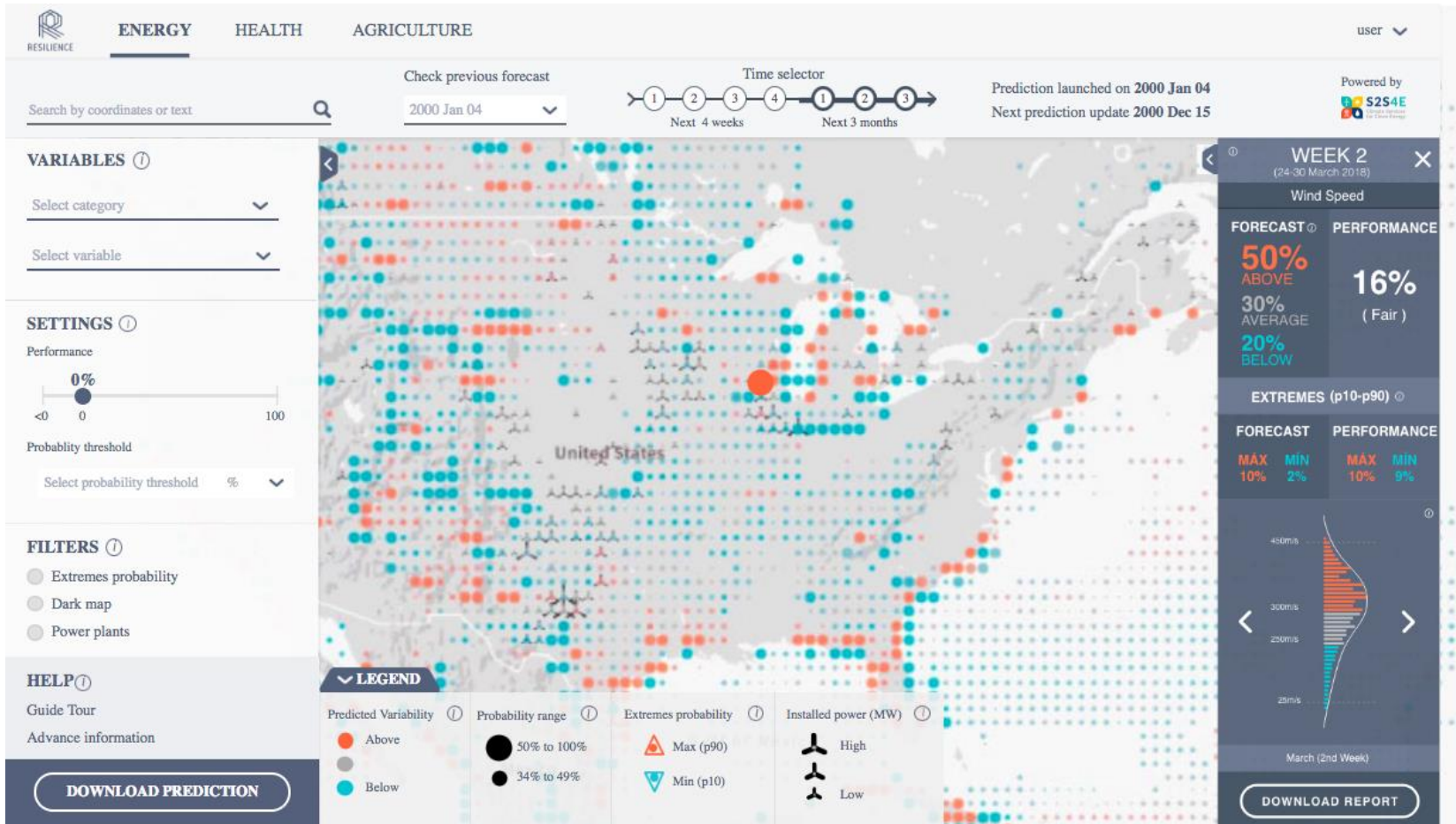


<http://www.bsc.es/ess/resilience/map.html>

DST, user feedback



Success criteria is TRUST



<http://www.bsc.es/ess/resilience/map.html>

Final remarks

- ▶ Climate prediction systems have improved in the last decade demonstrating that probabilistic forecasting can inform better decision making at some temporal scales and regions
- ▶ Alongside the model development process, climate predictions need to be evaluated on past years to provide robust information before making decisions
- ▶ Tailored service helpful for several applications
- ▶ Interdisciplinary groups enhance the interaction with users to co-develop a service

Future work:

- ▶ multi-model ensembles
- ▶ to improve the utility of forecasts by incorporating skillful information of the large-scale teleconnection patterns at different time scales

Thank you
Get in touch for more
information!



S2S4E

Climate Services
for Clean Energy

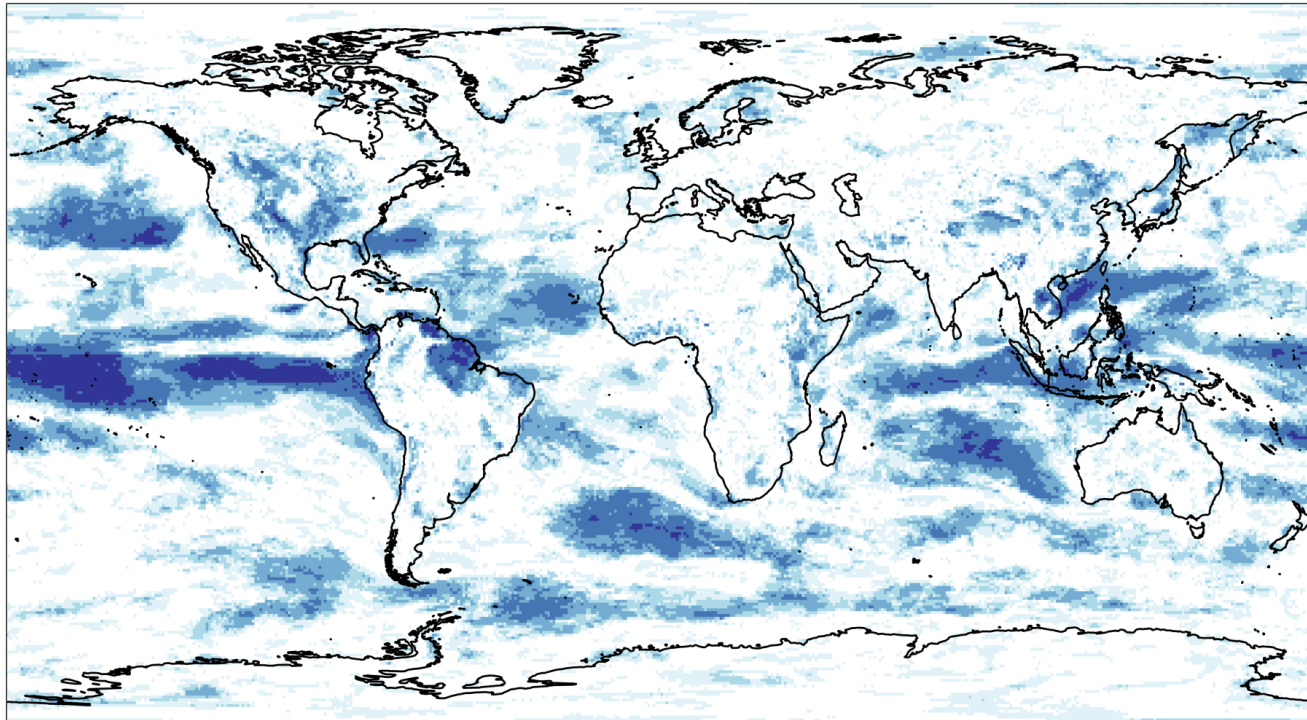
Albert Soret Miravet: albert.soret@bsc.es www.s2s4e.eu



Consortium

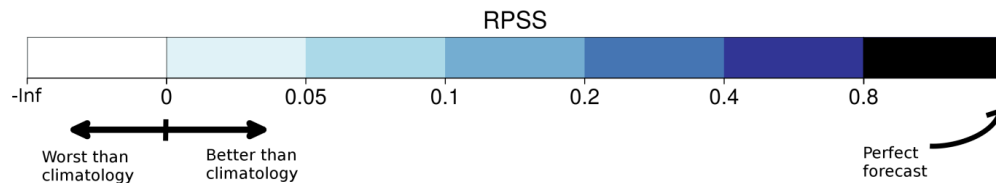


Skill



Skill assessment
for DJF (1981-2013)

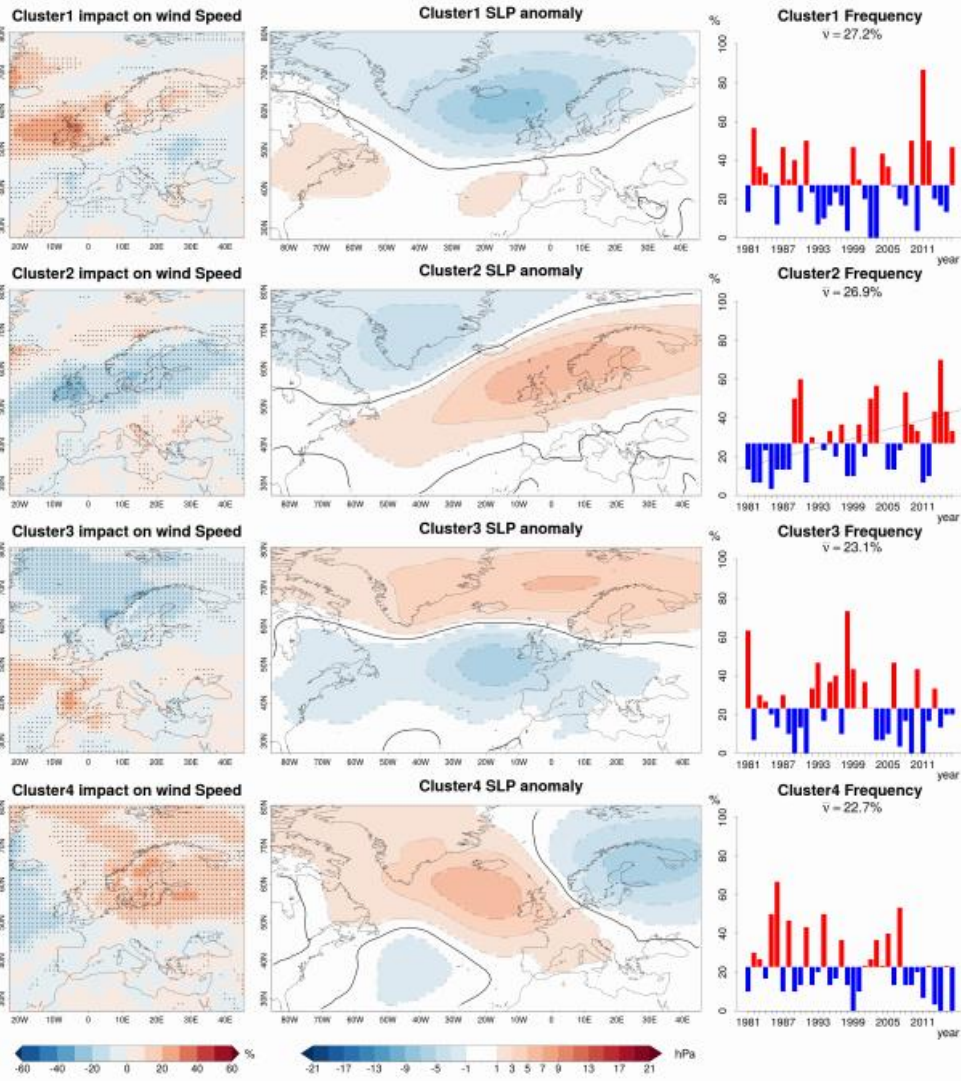
Displaying: Ranked
Probability Skill Score
[RPSS]



“A prediction has no value without an estimate of forecasting skill based on past performance”

Causes

JRA-55 / 10m wind speed and sea level pressure / Monthly anomalies and frequencies
September / 1981-2016



Center column: monthly SLP anomalies (in hPa) corresponding to the four Euro-Atlantic clusters (weather regimes) in September over the period 1981-2016, in decreasing order of explained variance.

Right column: monthly frequency of occurrence of the four clusters in September for 1981-2016. Eventual presence of black lines indicate significant trends.

Left column: impact of the four clusters on 10-m wind speed. Impact (in %) is relative to the average wind speed for the month of September over the period 1981-2016. Black dots indicate significant points with a t-test at 95% confidence level. (Source: JRA-55 reanalysis)

Region. Left: Europe (26.9°N-80.6°N, 23.1°W-45.6°E). Right: North Atlantic (26.9°N-80.6°N, 86.1°W-45.6°E)
Reference dataset: JRA-55 reanalysis

Capacity factor

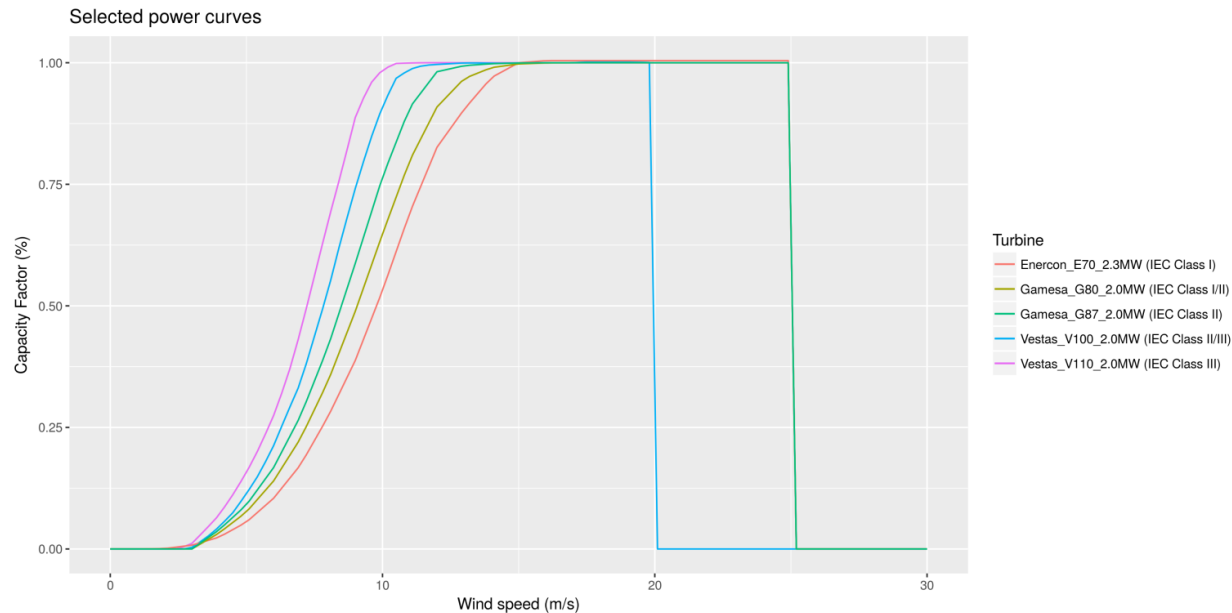
Capacity factor is a good indicator of wind power generation.

Is independent of:

- number of installed turbines
- nameplate capacity of installed turbines

Using manufacturer power curves for three turbines representing IEC classes.

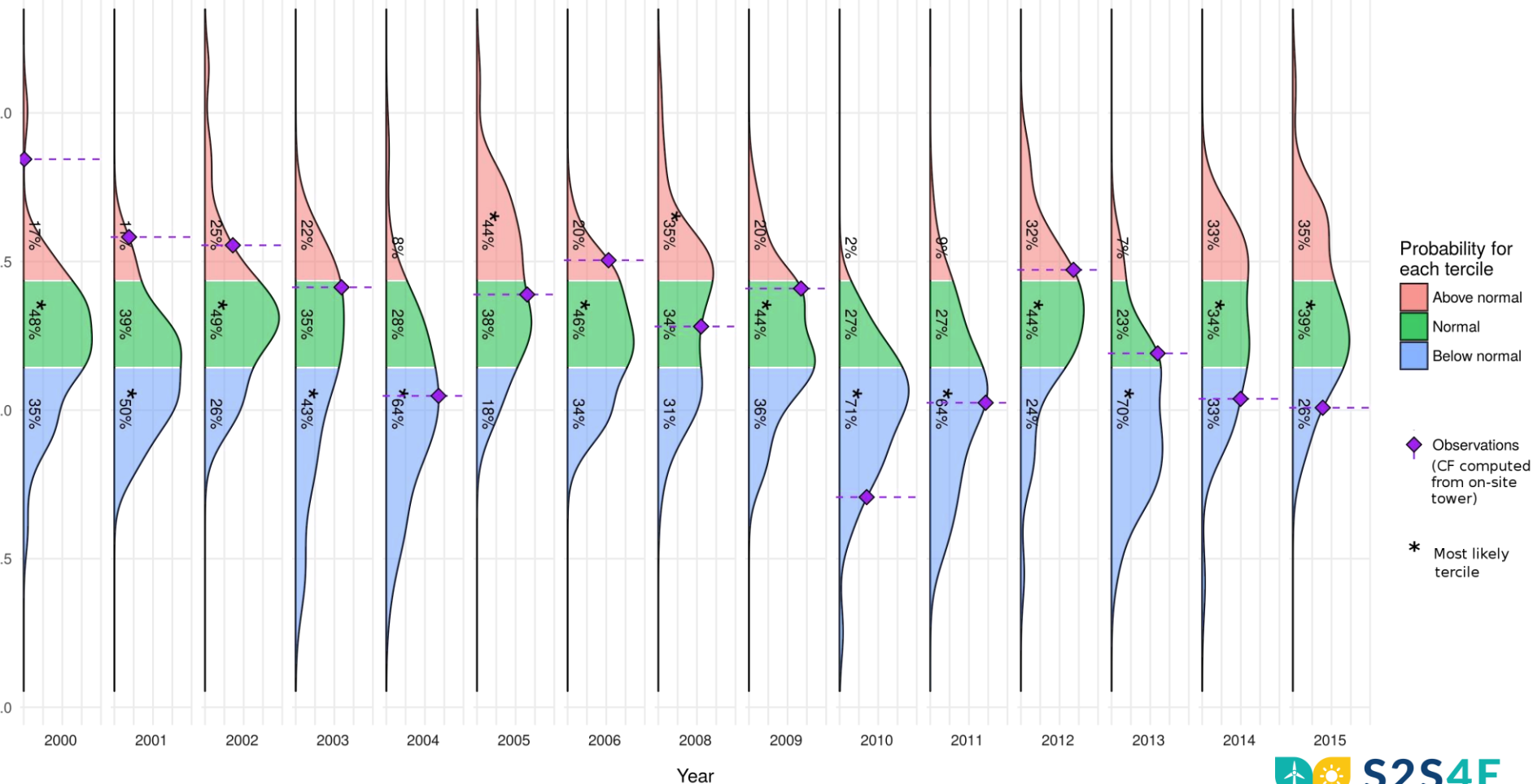
Fed with: 6-hourly model data, sheared at 100m.



Capacity factor

Retrospective forecasts for JJA at Site1 (2000-2015)

CRPSS = -0.01
 RPSS = 0.08
 EnsCorr = 0.44



Consortium

Society: Public

Co-operative

Research centers
Co-developing climate services



°CICERO
SMHI

Research centers
Co-developing climate services



SMEs
Data integration
the climate data factory
Cost analysis/
Business model
LGI
Forecaster
Nnergix

Industrial
Research info into
Decision process
EDF
renováveis
EnBW
Prototype DST
Capgemini

Industrial
Research info into
Decision process
EDF
renováveis
EnBW

Prototype DST
Capgemini
CONSULTING. TECHNOLOGY. OUTSOURCING

SMEs

Data integration
the climate data factory

Cost analysis/
Business model
LGI
sustainable innovation

Forecaster
Nnergix
Energy Forecasting Services

IBERDROLA
RENOVABLES

gasNatural
fenosa

VATTENFALL

VATTENREGLERINGSFÖRETAGEN
UMSÄLVEN • ÅNGERMANLÄVLEN • INDALÄLVLEN • LJUNGAN • LJUSMAN • DALÄLVLEN

Rte
TEM OPERATOR

Transmission system
operators (TSO)

S2S4E
Climate Services
for Clean Energy

Energy companies

Forecasters
ConWX



Synergies with other projects

SPECS and EUPORIAS

NEWA

CLIM4ENERGY and ECEM

SWICCA

QA4SEAS

INDECIS

MEDSCOPE

CLIM2POWER

ERA4CS

S2S

ClimatEurope

IMPRES

MARCO

VISCA and MEDGOLD

CLARA

SECLI FIRM

S2S4E

2017

2018

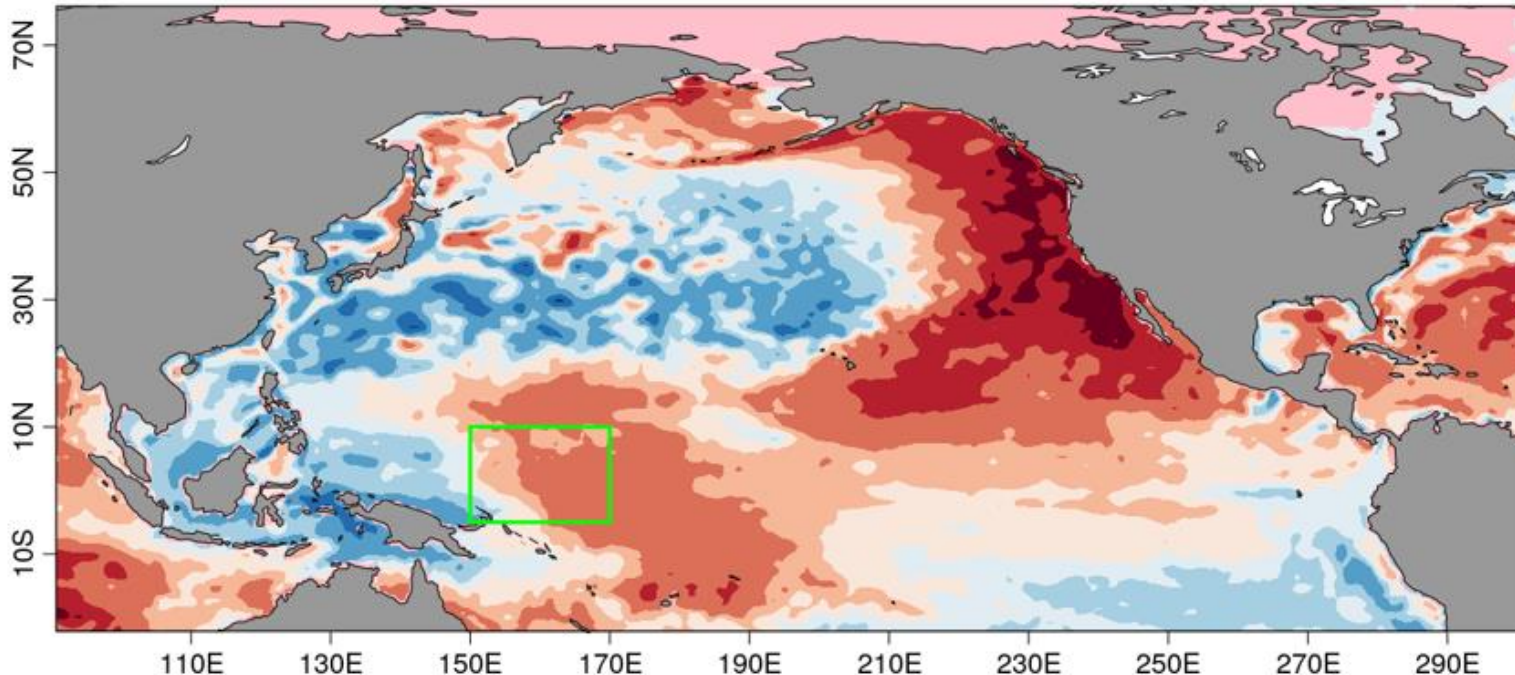
2019



Why?

Using retrospective climate predictions, we find that high ocean temperatures in the western tropical Pacific Ocean played a central role to establish and maintain those wind anomalies. This is not a single event. This work shows that the wind speed variability in the United States is not only dominated by El Niño but also by the ocean temperatures in this region of the Pacific.

Standardized SST anomalies for Q1 2015



Sea surface temperature anomalies in the Pacific Ocean during the same period. The green box shows the area under study.

