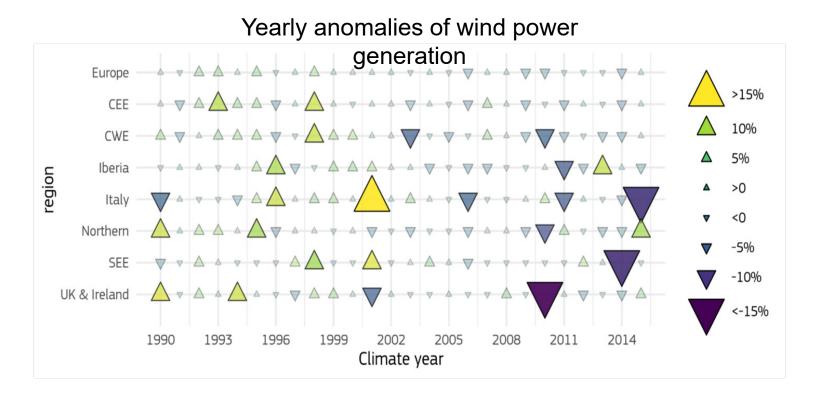


## Sub-seasonal to Seasonal Prediction for Energy

Frédéric Vitart, ECMWF Llorenc Lledo

EA State of the Art in Energy Forecasting Workshop, September 2022

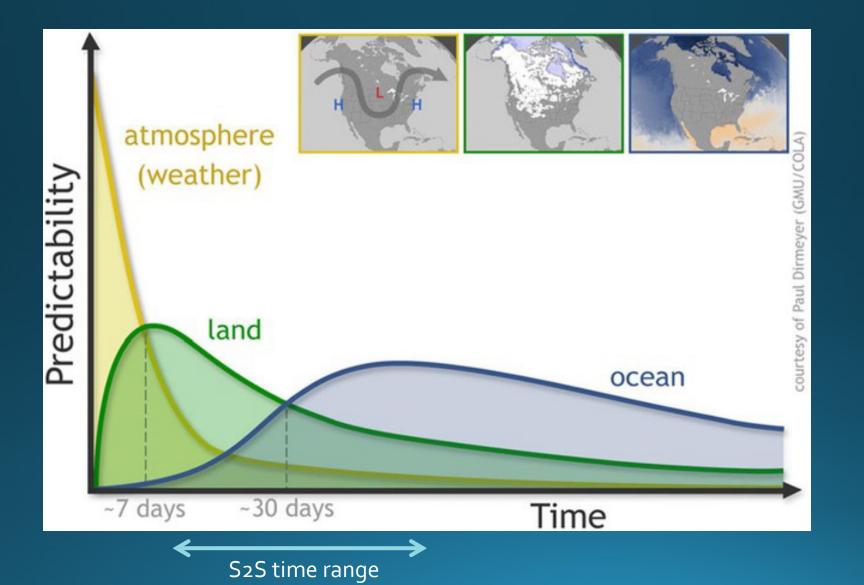
# With higher shares of renewables in the mix, the electricity system is more exposed to atmospheric variability risks



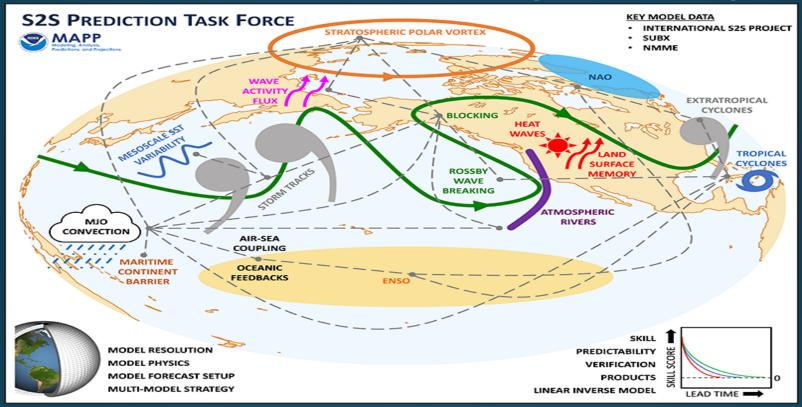


Can we anticipate weekly, monthly and seasonal anomalies of wind speed, wind power generation or energy demand?

## Extended-range predictability



## Sources of sub-seasonal and seasonal predictability

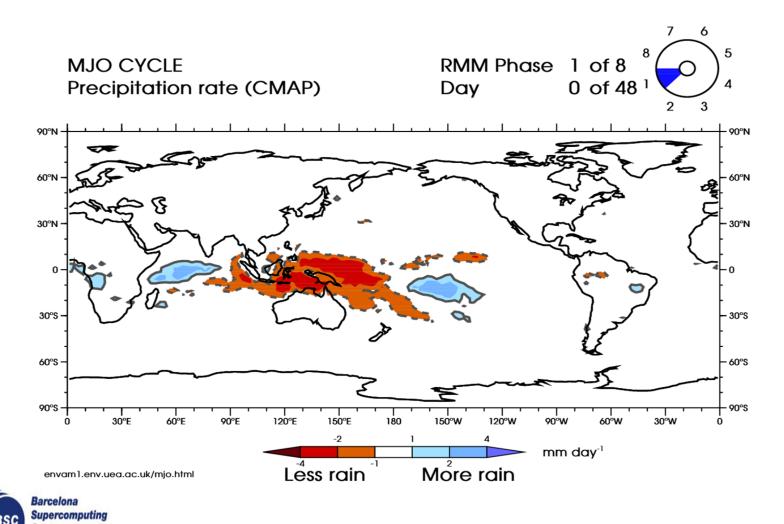


#### Main sources of predictability include:

- ENSOMadden Julian Oscillation
  - Soil Moisture
  - Stratospheric Initial conditions
  - Rossby waves
  - SSTs/Sea-ice
  - ≻Aerosols
  - ➢Others? …

Mariotti et al., 2019

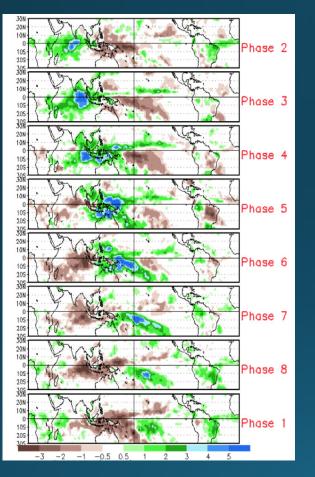
## The Madden-Julian Oscillation is the main source of subseasonal variability in the tropics



Center Centro Nacional de Supercomputación

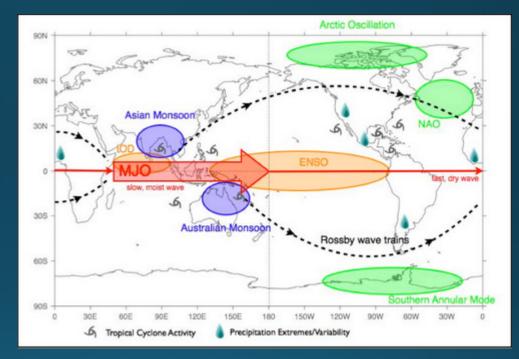
## The Madden Julian Oscillation

#### MJO phases

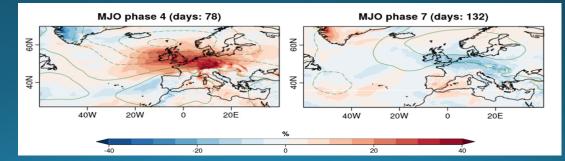


#### Gottschalk, 2014

#### **MJO** Impacts

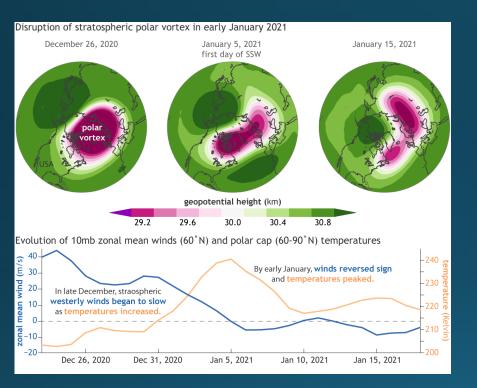


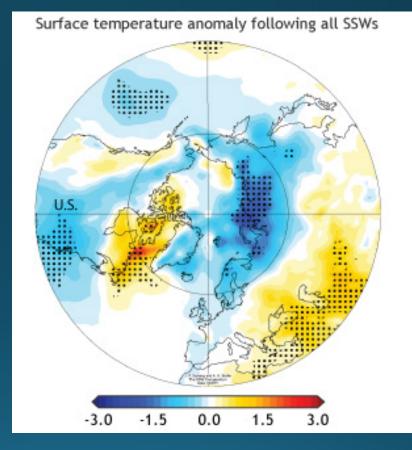
#### JFM surface wind anomalies in Europe



## Stratosphere: Weak vortex events

#### Jan 2021 weak vortex event





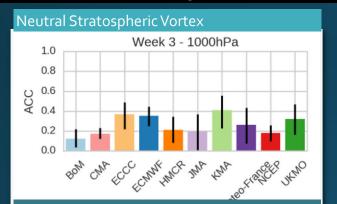
Surface temperature anomalies in degrees Celsius for (left) the 30 days following all identified SSWs in the NCEP-NCAR reanalysis record from 1958-2013. From <u>Butler et al. 2017</u>

#### L'Heureux 2021

## S<sub>2</sub>S prediction is challenging

Forecast skill is not constant in time. It depends strongly on the occurrence of sources of predictability: "windows of opportunity for forecast skill" (e.g. strong MJO, weak vortex event...)

Prediction skill of the 1000 hPa Northern Annular Mode for week 3 in the S2S models



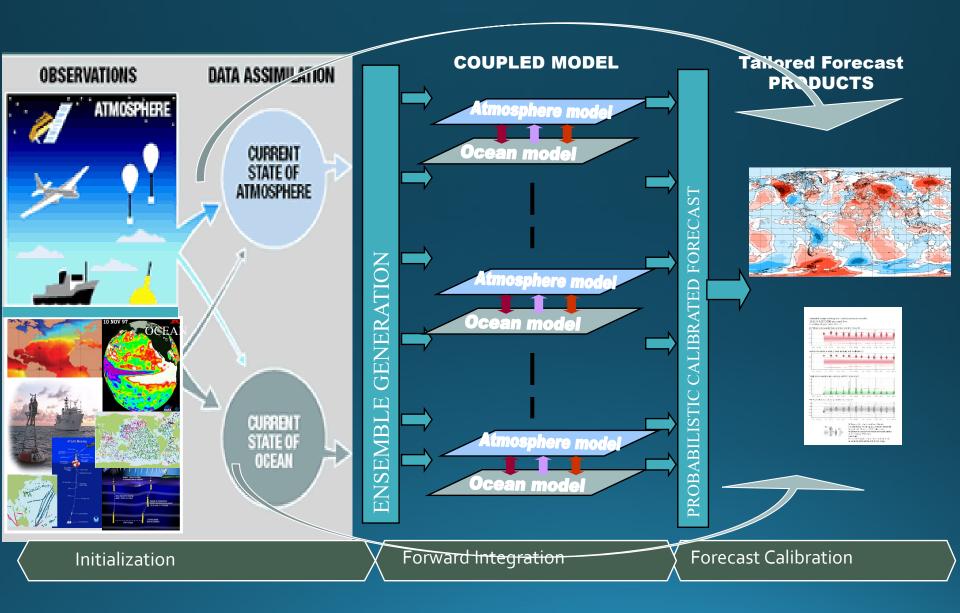
Domeisen et al. (2019)

• For most models, skill is higher following weak vortex conditions.

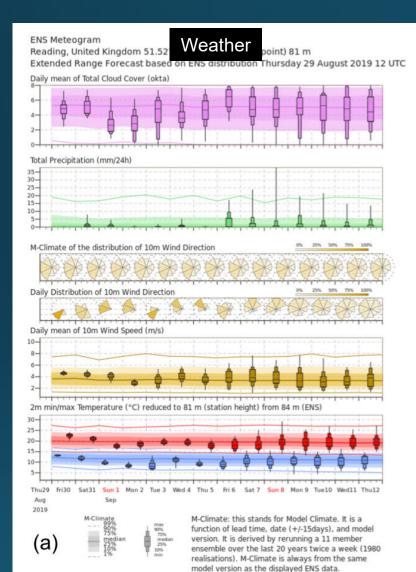
• Similar results are found following strong vortex conditions.

## How are S<sub>2</sub>S forecasts produced?

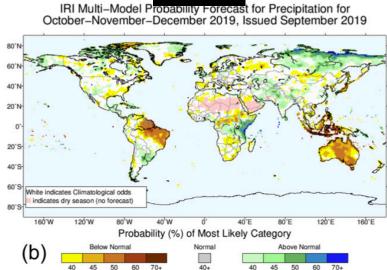
### End-To-End forecasting System

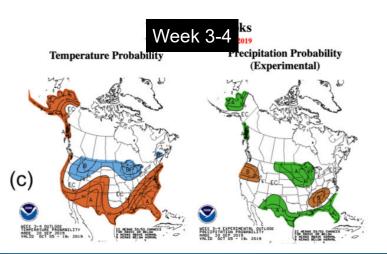


## What is the appropriate forecast format?



#### Seasonal

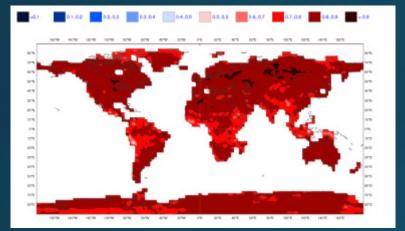




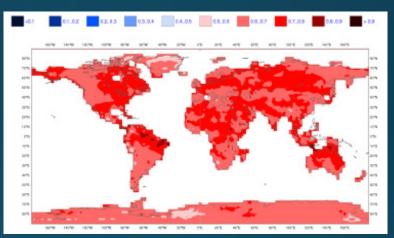
## What is the skill of 2-metre temperature?

ROC score: 2-meter temperature in the upper tercile

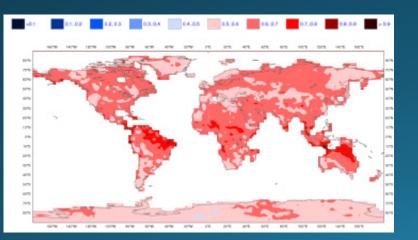
#### Day 5-11



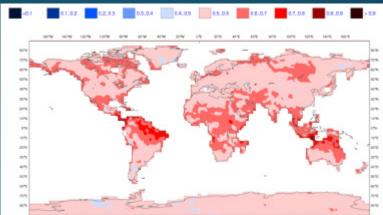
#### Day 12-18



#### Day 19-25



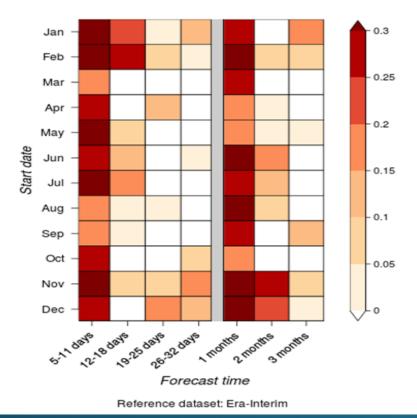
#### Day 26-32



NUME INCHE 180M 180M 181M 80M 40M 40M 20M 0/1 20M 40M 80M 80M 180M 180M 180M

# S2S skill evolution over the year

#### FairRPSS of ECMWF 10-m wind speed for 1996-2015 over Europe





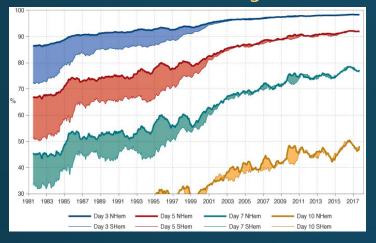
New EUROPEAN WIND ATLAS

#### Forecast skill characterization in the Fino 1 area for different start dates (Y axis) and forecast window (X axis).

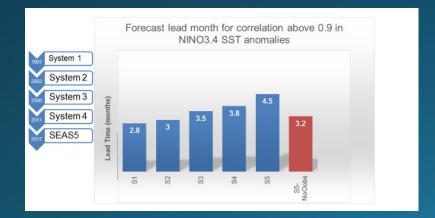


## Forecast skill. Are we filling the gap?

Weather forecasting

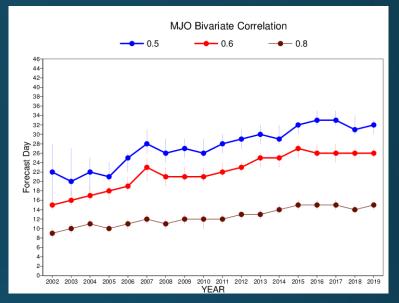


#### Toth and Buizza, 2018 Incremental Improvement => "quite revolution" Seasonal forecasting



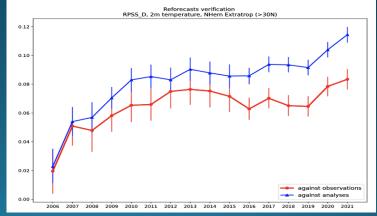
Improved ENSO prediction over past 20 years

#### Extended-range forecasting



#### Week 3 2m temp over N. Extratropics

#### Extended range: evaluation of re-forecasts

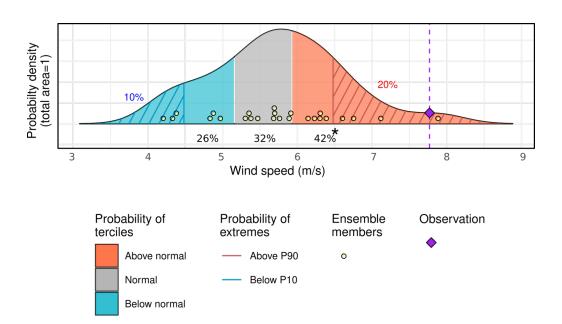


#### Rapid progress!

# CASE STUDIES from S2S4E

# A climate forcing can be detected in an ensemble of numerical simulations

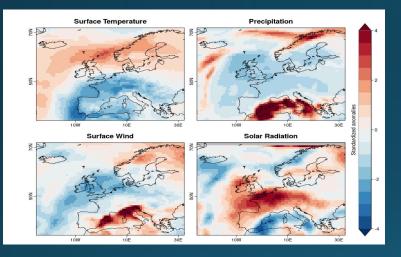
- Coupled Earth System Model Forecast fast & slow evolution fields and its interactions
- Ensemble members
  Each member represents one possible system evolution
- Signal extraction Average whole period to filter noise and obtain forcing signal
- Probabilities
  Count members above/below threshold



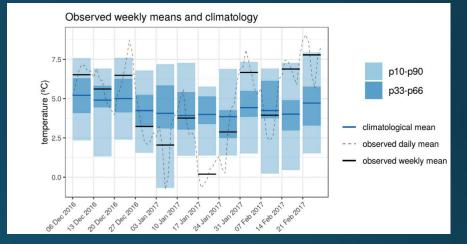


## Cold spell and wind drought in Europe January 2017

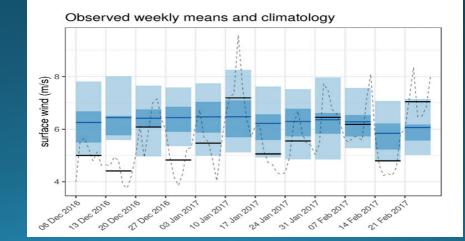
#### Standardized anomalies - Dec 2016-Feb 2017- ERA-Interim



#### Surface temperature



#### Wind Speed



#### Cold spell and wind drought in Europe January 2017 17 January 2017

#### Surface temperature forecast in Europe

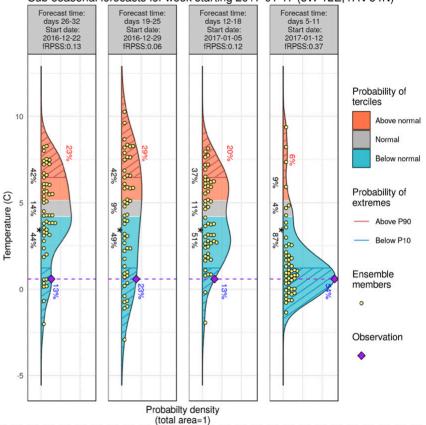
Electricity	y d	ema	nd f	orec	ast in	France

	Forecast lead time			
<b>Skill</b> (Temperature)	Days 26-32	Days 19-25	Days 12-18	Days 5-11
RPSS	0.13	0.06	0.12	0.37

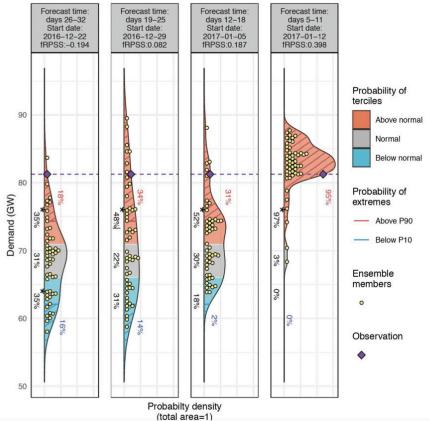
(see Bloomfield et al, 2021)

Skill	Forecast lead time			
(Electricity demand)	Days 26-32	Days 19-25	Days 12-18	Days 5-11
RPSS	-0.19	0.08	0.19	0.40

#### Sub-seasonal forecasts for week starting 2017-01-17 (5W-12E,47N-54N)

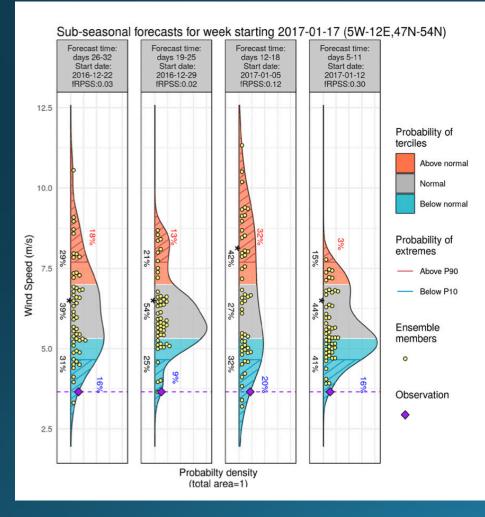






## Cold spell and wind drought in Europe January 2017 17 January 2017

#### Wind speed forecast

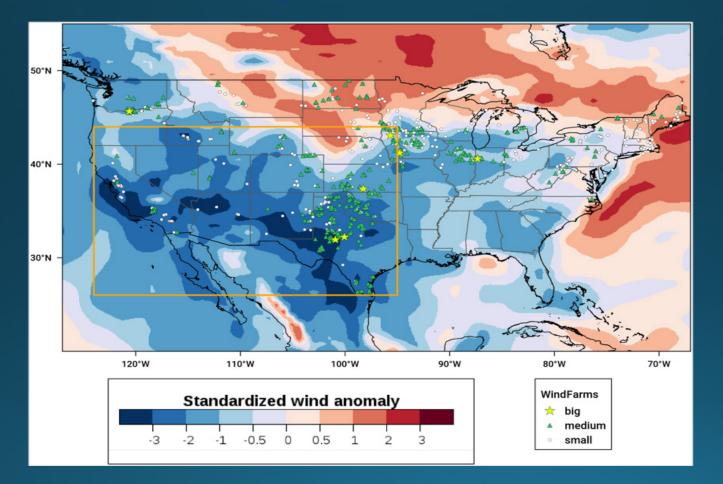


	Forecast lead time			
<b>Skill</b> (Wind Speed)	Days 26-32	Days 19-25	Days 12-18	Days 5-11
RPSS	0.03	0.02	0.12	0.30
BSS P10	0.07	-0.02	0	0.21
BSS P90	0.02	0.01	-0.01	0.19

# US wind drought

Q1 2015

## Wind anomaly Q1 2015





## **Event impacts**

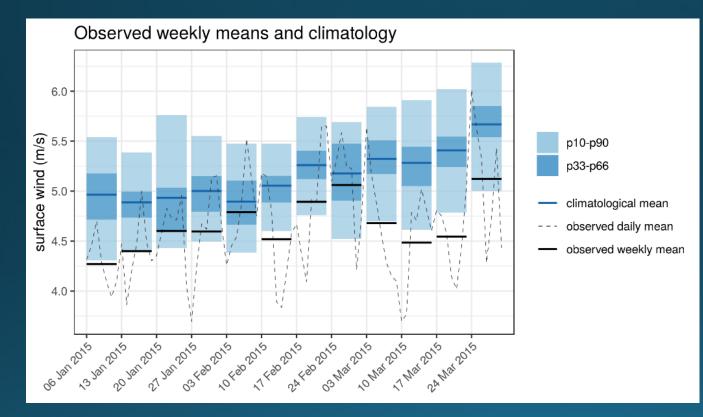


"US clean energy suffers from lack of wind" *Financial Times, September 2015.* 

"We never anticipated a drop-off in the wind resource as we have witnessed over the past six months" *David Crane, RNG, September 2015.* 



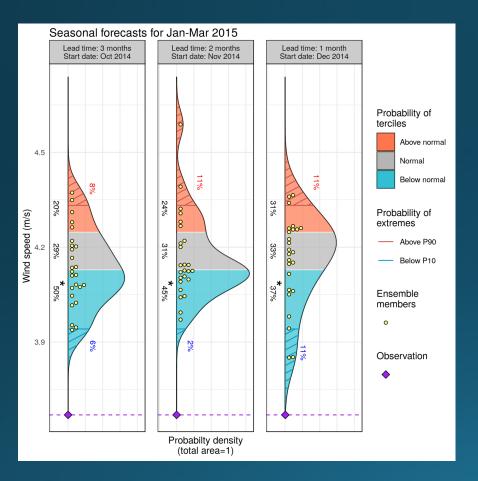
## Widespread and extended in time







## Fcsts available 3 to 1 months ahead



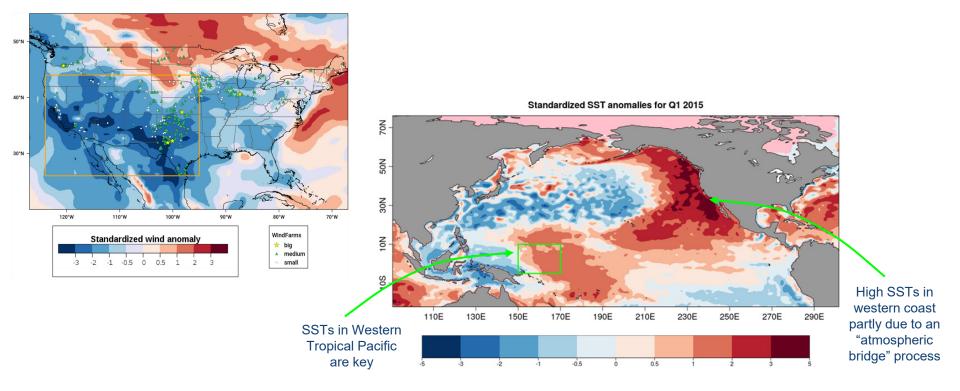
Associated Skill Scores				
	Start Date			
	Oct	Nov	Dec	
RPSS	0.35	0.39	0.35	
BS P10	-0.07	-0.27	-0.16	
BS P90	0.1	0.04	0.07	
CRPSS	0.14	0.11	0.14	
EnsCorr	0.55	0.54	0.51	

Which decisions would you make in view of those forecasts?

System: ECMWF SEAS5 Reanalysis: ERA-Interim Bias adjustment: calibration Hindcast: 1993-2015

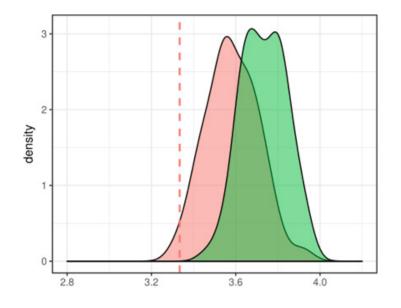


# Why did it happen?





# **Attribution experiment with EC-EARTH**



2 experiments w 100 members initialized 1st January 2015



**PDF(wind | normal SST)** 

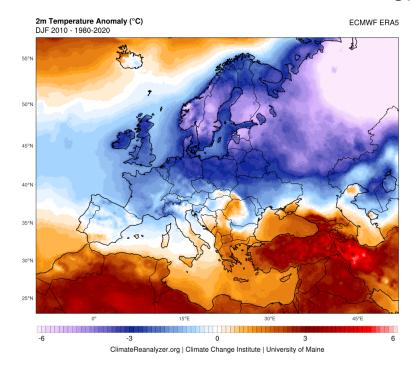


PDF(wind | high SST)

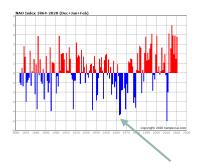


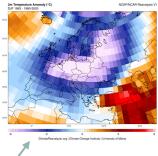
## A NAO<sup>—</sup> as in 2009/10 winter would freeze Europe's most vulnerable homes

2009/2010 winter was cold and still: energy demand increased up to 20%



- Sea level pressure difference between Iceland and Açores very weak (NAO<sup>-</sup> pattern)
- Easterly flow brings cold continental air
- 1962/63 is an analogue year based on NAO index

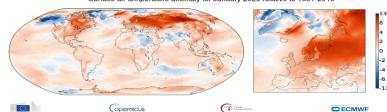




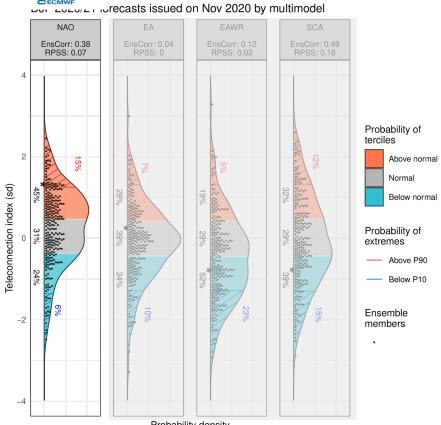


#### January 2020: strong NAO+ pattern perature anomaly for January 2020 relative

°C



- Multi-system predictions of 4 EATCs •
- Combine info from several sources
- Skill almost as good a best system, and more robust.
- Extremes (PoE P10/P90) are better described with • more members.
- Summary ٠
  - Enhanced prob. of NAO+ 0
  - Enhanced prob. of SCA-0
  - High prob. of EAWR-0







## Conclusions

 S2S prediction is challenging. Skill at this time range depends strongly on the occurrence of sources of predictability

 S2S models are improving with better representation and prediction of sources of predictability such as the MJO and stratospheric processes

 There is room for improvement through the use of AI/ML methods to better calibrate model outputs.

• Case studies suggest that S2S forecasts can be useful for the prediction od energy demand. However, S2S prediction for wind production is more challenging.