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# Seasonal Forecasting for the European energy sector

C3S European Climatic Energy Mixes (ECEM) Webinar 18<sup>th</sup> Oct 2017

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### Seasonal Climate Prediction

- Forecasting for several months (e.g. a 3-month season)
   with a lead time of several weeks (e.g. 1 month) → e.g. forecasting for DJF in Oct/Nov
- Traditional weather forecasts lose predictability after around 10 days, due to the naturally chaotic nature of the atmosphere...
- But some parts of the climate system change slower than the atmosphere, e.g.
  - Oceans
  - Sea ice
  - Soil moisture
- For example the ocean is still predictable even when the atmosphere is not!
   → we can predict how the atmosphere might respond to the oceans.











### Seasonal Climate Prediction

- Forecasting for several months (e.g. a 3-month season)
   with a lead time of several weeks (e.g. 1 month) → e.g. forecasting for DJF in Oct/Nov
- So we use a coupled climate model (atmosphere + land + ocean + sea ice), initialised with current observations, and try to forecast the climate for the coming season....
- Climate drivers are large-scale phenomena, so forecasts are only skillful over large areas (e.g. Spain-sized!), and are probabilistic.
- Seasonal climate predictions are *not* "extended weather forecasts"!
  - We can't predict the weather on 25<sup>th</sup> January from 1<sup>st</sup> Nov forecast!
  - We can estimate how likely the Dec-Jan-Feb period is to be colder/warmer than normal (if skillful).
- → Seasonal forecasts are useful for different types of decisions (long-term risk-based planning)
- Use as **part** of a decision-making chain, along with monthly, weekly & daily forecasts.





#### Seasonal Forecasting in ECEM

- Demonstrate the use of seasonal forecast data from C3S for the energy sector, linking with data sets produced elsewhere in the project.
- Before forecasts can be provided, the skill of the forecasting systems must be assessed.
- Skill depends on the variable in question, the particular model used, and the region and season being investigated.
- Assess the skill of climate variables relevant to the energy sector
   → Data now on Demonstrator, full results in report → final update Nov 2017
- Assess the skill of methods of forecasting energy variables themselves (demand, wind power, etc.), for select countries.
   → Due: December 2017

Additional work: **Case studies** (Emma Suckling, U. Reading) → See previous webinar



### Seasonal Forecast Skill

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The skill of the forecast system is estimated from a series of re-forecasts over a historical period (hindcasts), compared to observations over that period.

Hindcast data sets are limited (number of years, number of realizations each year), and the observations can be uncertain – so skill estimates are also uncertain!

Real forecasts are often produced using a slightly different system to the hindcast (e.g. more realizations), so can be more accurate than the skill assessment suggests.

European



#### Models and Parameters



#### **Seasonal forecasting systems:**

Originator	Forecast system	Model	Spatial resolution (approx)	Hindcast period	Hindcast ensemble size	Forecast ensemble size
ECMWF	System 4	IFS Cyc36r4	80 km	1981–2010 (30 yrs)	51	51
Météo-France	System 5	Arpege-IFS Cyc37	80 km	1993–2014 (22 yrs)	15	51
Met Office	GloSea5	HadGEM3-GC2	60 km	1993–2015 (23 yrs)	28	42

#### Variables:

- Seasonal mean 10m wind speed (m/s)
- Seasonal mean 2m air temperature (°C)
- Seasonal mean 2m relative humidity (%)
- Seasonal total precipitation amount (mm)
- Seasonal mean downwelling shortwave irradiance (W/m<sup>2</sup>)
- Seasonal mean air pressure at sea level (hPa)

#### Seasons:

- Winter (DJF)
- Summer (JJA)

#### **Observational data:**

- ERA-Interim
- ECEM bias-adjusted reanalysis

#### **Skill statistics:**

#### Gridded maps:

Correlations

#### *Country-average:*

- Timeseries & correlations
- Reliability diagrams & Brier skill scores
- ROC diagrams & ROC skill scores





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#### Maps of correlation skill, against ERA-Interim:



Skill is not universal,

but in this case there are some common signals across the models.

Note: We're only assessing direct model skill here. In some cases, better results can be obtained using larger-scale drivers (e.g. NAO) as the predictor of temperature etc.



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#### Maps of correlation skill, against ECEM bias-adjusted reanalysis:



At the seasonal time scale, there is very little change in the correlation skill when moving to the ECEM bias-adjusted reanalysis



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#### Correlation skill in countries:



Getting results for individual countries is essential, but care needs to be taken with these results.

We're taking a very simple approach to country-averaging, and comparing against the gridded data can be helpful.



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Romania shows significant skill across models for JJA temperatures.

There is a trend in the observations, and the ability of the forecast systems to reproduce the trend contributes to the positive skill.

("Standardised" or "normalised": Plotting  $(T - \mu_T) / \sigma_T$  instead of just T here, i.e. removing the mean and scaling by the variability, to match the calculation of the correlation)



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## Summary: JJA Correlations





-0.4 -0.2 0.0 0.2 0.4 0.6 0.8

Pearson correlation for JJA temperature

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-0.6 -0.4 -0.2 0.0 0.2 0.4 0.6 0.8 -1.0 -0.8 Pearson correlation for JJA accumulated precipitation



-0.6 -0.4 -0.2 0.0 0.2 0.4 0.6 0.8 -1.0 -0.8 Pearson correlation for JJA irradiance



-0.6 -0.4 -0.2 0.0 0.2 0.4 0.6 0.8 -1.0 -0.8



**GloSea5** Office Met

-0.6

-0.6 -0.4 -0.2 0.0 0.2 0.4 0.6 0.8 1.0 Pearson correlation for JJA temperature



-0.6 -0.4 -0.2 0.0 0.2 0.4 0.6 0.8 1.0 -1.0 -0.8 Pearson correlation for JJA wind speed at 10m



-1.0 -0.8 -0.6 -0.4 -0.2 0.0 0.2 0.4 0.6 0.8 Pearson correlation for JJA accumulated precipitation



-1.0 -0.8 -0.6 -0.4 -0.2 0.0 0.2 0.4 0.6 0.8 1.0 Pearson correlation for JJA irradiance



### Summary: **DJF** Correlations

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 Met Office Gloses

 Available Nov 2017



#### The diversity of skill

Count	Country		Met Office					WF				Météo-France					
Code	Name	WS	T2M	RH	TP	SSRD	WS	T2M	RH	TP	SSRD	WS	T2M	RH	TP	SSRD	
AL	Albania	C						C									
AT	Austria												C-R				
BE	Belgium									C							
BA	Bosnia-		C					C									
	Herzegovina																
BG	Bulgaria		C					C-R	C				C-R				
HR	Croatia		C	C				C-R									
CZ	Czech				R								C				
	Republic																
DK	Denmark			C-R													
EE	Estonia																
FI	Finland							C									
FR	France							C-R									
DE	Germany																
GR	Greece		C					C-R									
HU	Hungary		C	C				C-R	CBR				C				
IE	Ireland																
IT	Italy							C-R									
LV	Latvia																
LT	Lithuania																
LU	Luxembourg																
МК	Macedonia	C						C									
ME	Montenegro		C					C					C				
NL	Netherlands																
NO	Norway							C-R									
PL	Poland						R										
PT	Portugal																
RO	Romania		CBR	C				CBR	C-R				C-R				
RS	Serbia		C					C									
SK	Slovakia							R	CBR								
SI	Slovenia							C									
ES	Spain							CBR	C								
SE	Sweden																
СН	Switzerland																
UK	United																
	Kingdom																

# JJA skill:

Where a skill score is significantly greater than zero, it is marked with a **C** (correlation), **B** (Brier skill score) or **R** (ROC skill score).

Colours: 1 score, 2 scores, 3 scores

Skill is diverse across models, variables and seasons.

Having more significant skill scores can add confidence, but the behaviour of the models should be examined in detail for each use case.





### The diversity of skill

Country			Met Office						WF			Météo-France					
Code	Name	I	WS	T2M	RH	TP	SSRD	WS	T2M	RH	TP	SSRD	WS	T2M	RH	TP	SSR
AL	Albania	Π	[							C		C					
AT	Austria	1							C					C			
BE	Belgium	1							C	C			C	C			
BA	Bosnia-	1										C					
	Herzegovina																
BG	Bulgaria																
HR	Croatia								R		C	C					
CZ	Czech								C	C		C		C			
	Republic																
DK	Denmark															C	
EE	Estonia				·			C	C								
FI	Finland							C-R	C							C	
FR	France					>								C			
DE	Germany								C	C-R			C	C	C-R		
GR	Greece									C	C	C					
HU	Hungary					1						C					
IE	Ireland								C	C						C-R	
IT	Italy	1			<u> </u>												
LV	Latvia	1											C				
LT	Lithuania	1			<b>B</b>	)							C				
LU	Luxembourg					L			C	C				C			
MK	Macedonia					<b>[</b> ]				C		C					
ME	Montenegro									C		C					
NL	Netherlands					L			C	C-R			C	C		C	
NO	Norway				<b>F</b>	<b>)</b>		C-R	C-R	R	C						
PL	Poland					I				C		R	C	C			
PT	Portugal				•												
RO	Romania				Т	<b></b>											
RS	Serbia				1.1							C					
SK	Slovakia											C		C			
SI	Slovenia								C					C			
ES	Spain																
SE	Sweden							C-	C-R							C	
СН	Switzerland												C	C			
UK	United								C-R								
	Kingdom																

# **DJF** skill:

Where a skill score is significantly greater than zero, it is marked with a **C** (correlation), **B** (Brier skill score) or **R** (ROC skill score).

Colours: 1 score, 2 scores, 3 scores

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Having more significant skill scores can add confidence, but the behaviour of the models should be examined in detail for each use case.



# SUMMARY

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- Seasonal forecasting uses **global climate models** to predict the climate of the coming 3-month season, with a lead time of about a month.
- Skill comes from slowly-varying components of the global climate system
- Uncertainty means we forecast probabilities of average seasonal conditions
  - e.g. "a 60% probability of the Dec-Jan-Feb period on average being colder than normal"
- We use hindcasts of the historical period to estimate the skill of the forecast systems. The limited size of the hindcasts makes these estimates uncertain.
- Skill in Europe is very diverse across different models, variables and seasons.
  - How skillful do **you** need a seasonal forecast to be, before it is useful for your decisions?
- Initial estimates of skill are provided in the ECEM Demonstrator, providing a starting point when considering if a seasonal forecast could be useful.





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# Thank you for listening!



