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# Stratospheric influences on subseasonal predictability of European energy-industry-relevant parameters

Dominik Büeler <sup>1,2</sup> / Remo Beerli <sup>3,2</sup>, Heini Wernli <sup>2</sup>, Christian M. Grams <sup>1,2</sup>

- 1) Institute of Meteorology and Climate Research, Department Troposphere Research, KIT, Germany
- 2) Institute for Atmospheric and Climate Science, ETH Zurich, Switzerland
- AXPO Solutions AG, Switzerland



### Motivation | Polar vortex – weather regimes – wind power

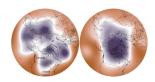
e.g., Baldwin & Dunkerton, 2001, SCI; e.g., Clark et al., 2017, ERL; Tripathi et al., 2015, ERL; Brayshaw et al., 2011, RE Charlton-Perez et al., 2018, QJRMS NOAA EnergyWay NOAA

State of the stratospheric polar vortex (SPV) as a direct source of subseasonal predictability for European energy industry?

Dominik Büeler | dominik.bueeler@kit.edu

# Data | Statistical forecast





- Strength of SPV
  - $\bullet$  ( $\Delta$  Z@150hPa)<sub>60°-90°N</sub> from ERA-Interim
  - Daily, DJF, 1985 2014



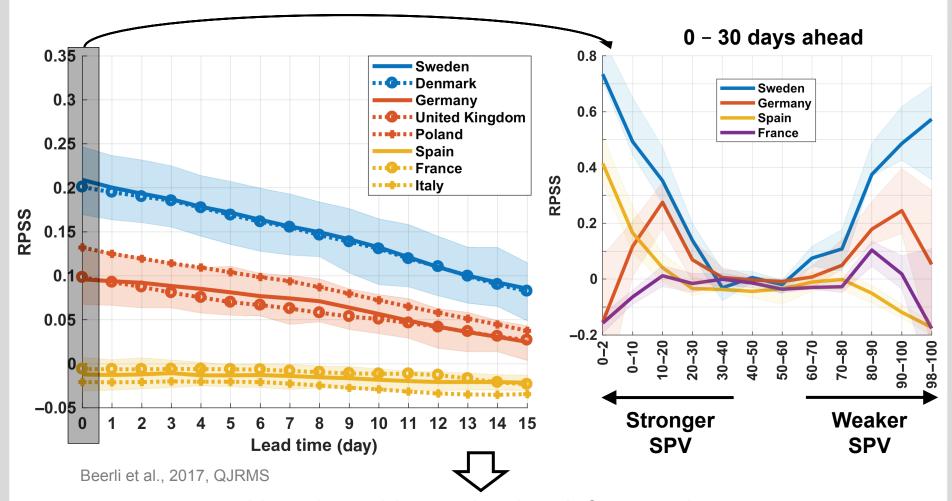
- Wind power generation for every European country
  - Renewables.ninja dataset (Staffel & Pfenninger, 2016, ENE; www.renewables.ninja)
  - Daily month-ahead average, DJF, 1985 2014

Beerli et al., 2017, QJRMS

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# Karlsruhe Institute of Technology

### Results | Simple 3-categorical statistical forecast



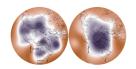
How does this mechanism influence the skill of subseasonal numerical weather models?

# Data | Numerical forecast



- Subseasonal ECMWF model (<u>www.s2sprediction.net</u>)
  - 2 reforecasts / week, DJF, 1995 2017
  - 11 ensemble members

#### Fields calculated for each reforecast



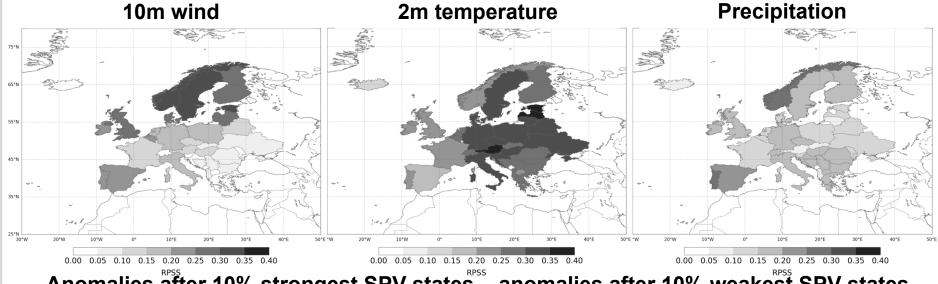
- Strength of SPV = (Δ Z@100hPa)<sub>60°-90°N</sub>
- At forecast initial time



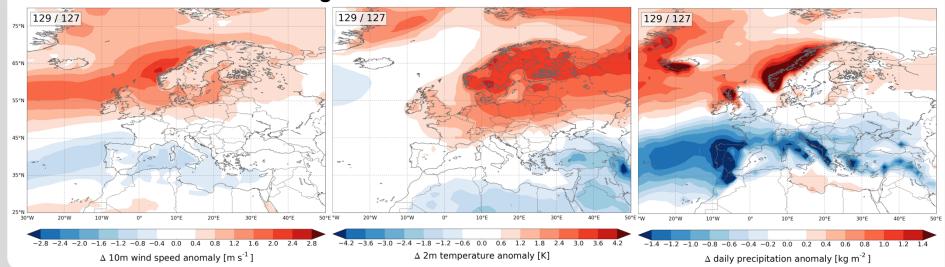
- (Δ 10m wind) <sub>European Countries</sub>
- (Δ 2m temperature) European Countries
- (Δ precipitation) <sub>European Countries</sub>
- Average over 1 month lead time

# Results | Regional model skill pattern



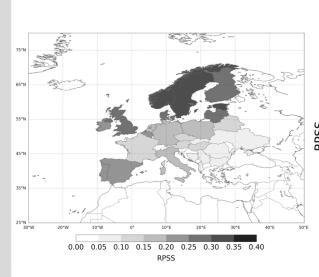


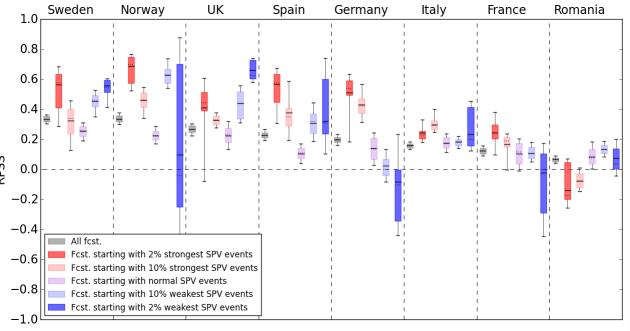
# Anomalies after 10% strongest SPV states anomalies after 10% weakest SPV states



# Results | Model skill for 10m wind



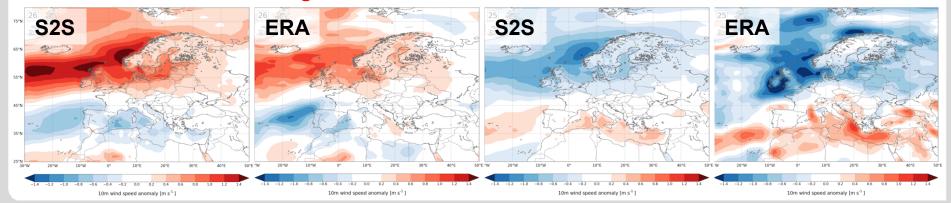




#### **Anomalies after 2% strongest SPV states**

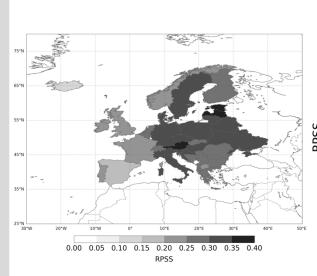
Dominik Büeler | dominik.bueeler@kit.edu

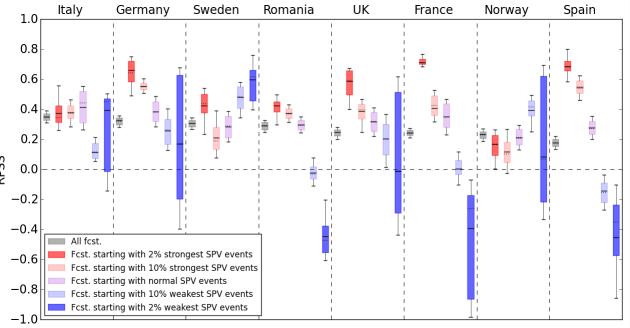
#### **Anomalies after 2% weakest SPV states**



# Results | Model skill for 2m temperature

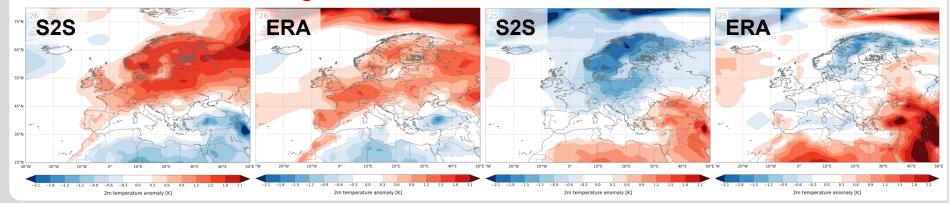






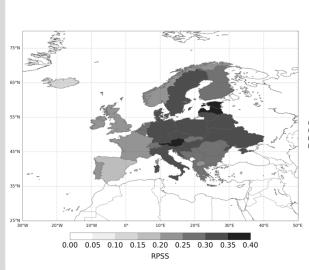
#### **Anomalies after 2% strongest SPV states**

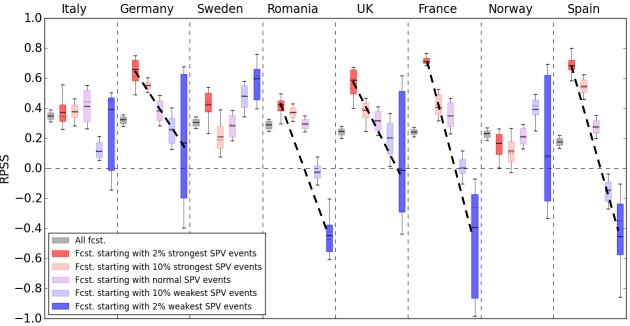
#### **Anomalies after 2% weakest SPV states**



# Results | Model skill for 2m temperature

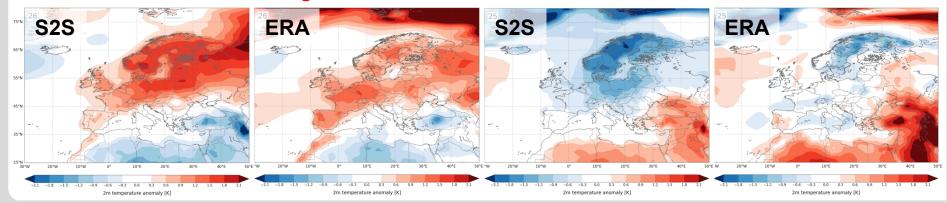






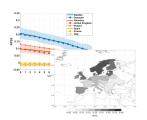
#### **Anomalies after 2% strongest SPV states**

#### **Anomalies after 2% weakest SPV states**

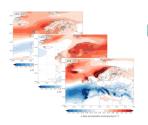


#### Conclusions





Strong spatial variability of statistical and numerical model skill for month-ahead prediction of wind power generation / surface weather in Europe



■ Reason: anomalous SPV states at forecast initial time lead to persistent NAO-like anomaly patterns → model skill for countries located in affected regions tends to be enhanced



However, model skill increase much more significant and robust after strongest SPV states than after weakest SPV states (~ SSWs), which even lead to significant skill reduction for certain countries (particularly T@2m)





### Implications

Dominik Büeler | dominik.bueeler@kit.edu

- Energy meteorology cannot rely on enhanced predictability after weakest (~ SSWs) but more after strongest SPV states
- Regional SSW response in S2S models needs to be improved