

Accuracy of near real time updates in wind power forecasting with regard to different weather regimes

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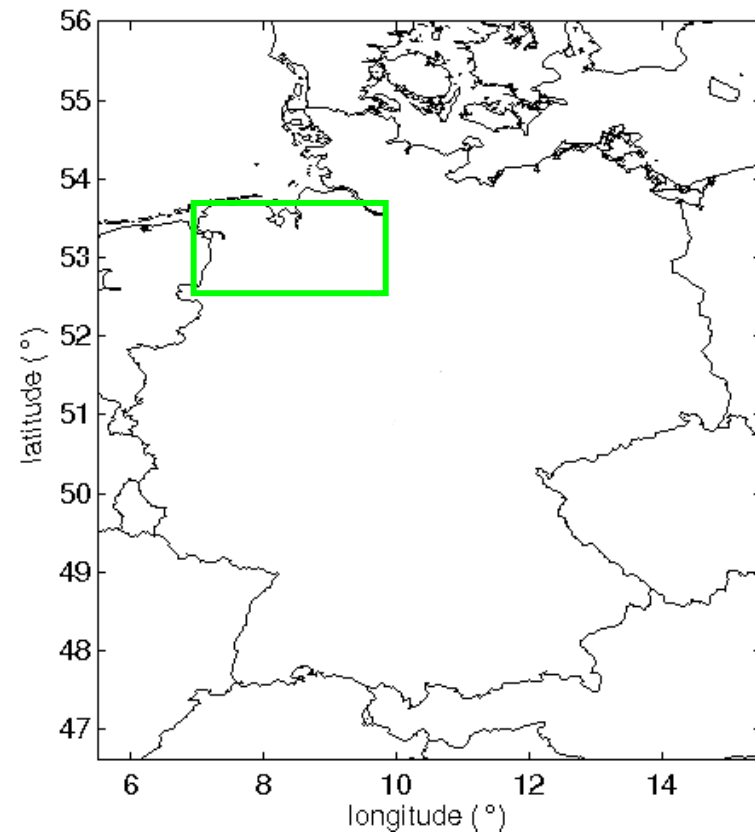
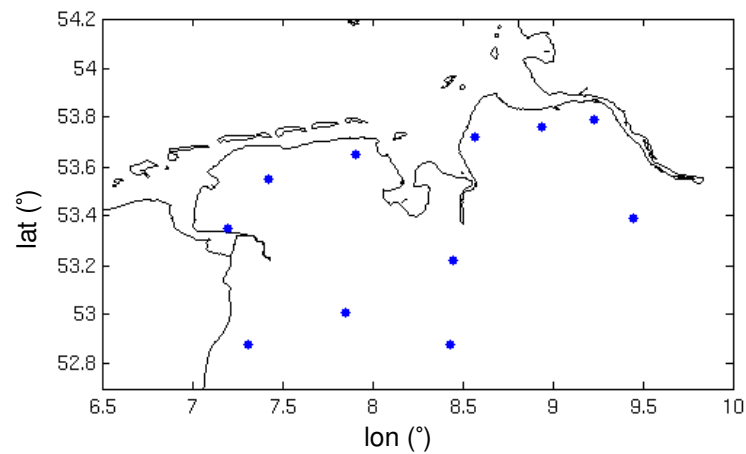
7th EMS Annual Meeting
8th European Conference on Applications of Meteorology
San Lorenzo de El Escorial, Spain, 01 – 05 October 2007

Outline

- Study site
- Wind power forecasting - method
- Cluster analysis – method and results
- Observed power by clusters
- Forecast errors by clusters
- Conclusions

Study site

North-West-Germany
single wind farms



Wind power forecast

→ data ←

observed wind power input (2004 – 2006)

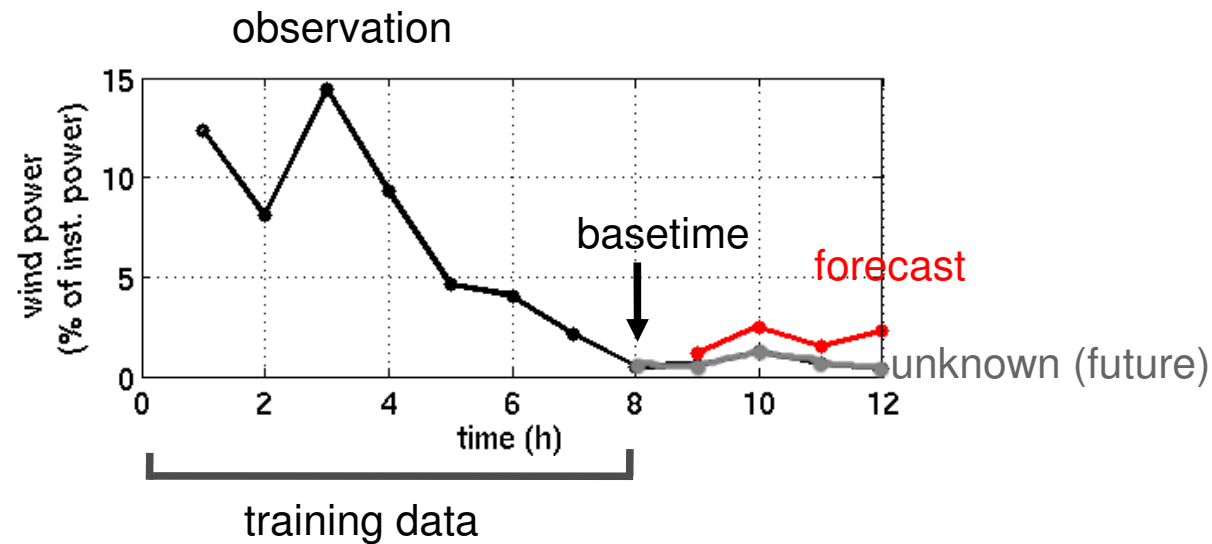
→ objective ←

forecast wind power of the next 4 hours

without wind speed information from weather forecasts (Numerical Weather Prediction)

→ method ←

Neural Networks



Clustering

data: 500 hPa heights from ECMWF analysis data (6-hourly), Jan. 2005 – April 2007

Principal Component Analysis (PCA)

- reduction of data
- take as much components to have 99 % of explained variance

→ relate single clusters to points in time

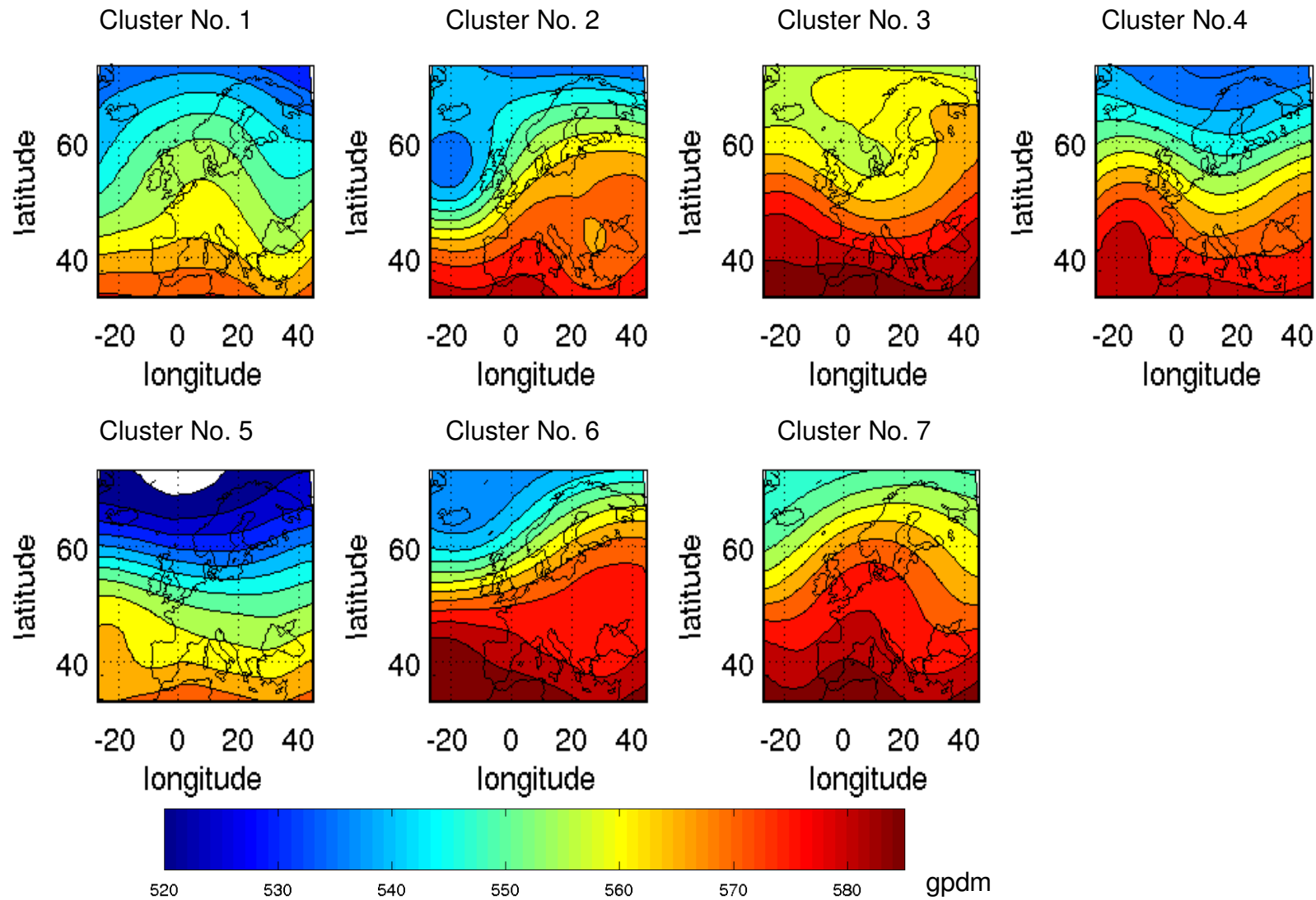
Cluster analysis

- k-mean clustering
- separately for:
summer (April - Sept) and
winter (Oct. - March)

time	cluster
03-May-2005 06:00:00	1
03-May-2005 12:00:00	1
03-May-2005 18:00:00	1
04-May-2005 00:00:00	1
04-May-2005 06:00:00	1
04-May-2005 12:00:00	4
04-May-2005 18:00:00	4
05-May-2005 00:00:00	5
05-May-2005 06:00:00	5
05-May-2005 12:00:00	5
05-May-2005 18:00:00	5
06-May-2005 00:00:00	5
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06-May-2005 12:00:00	5
06-May-2005 18:00:00	5
07-May-2005 00:00:00	5

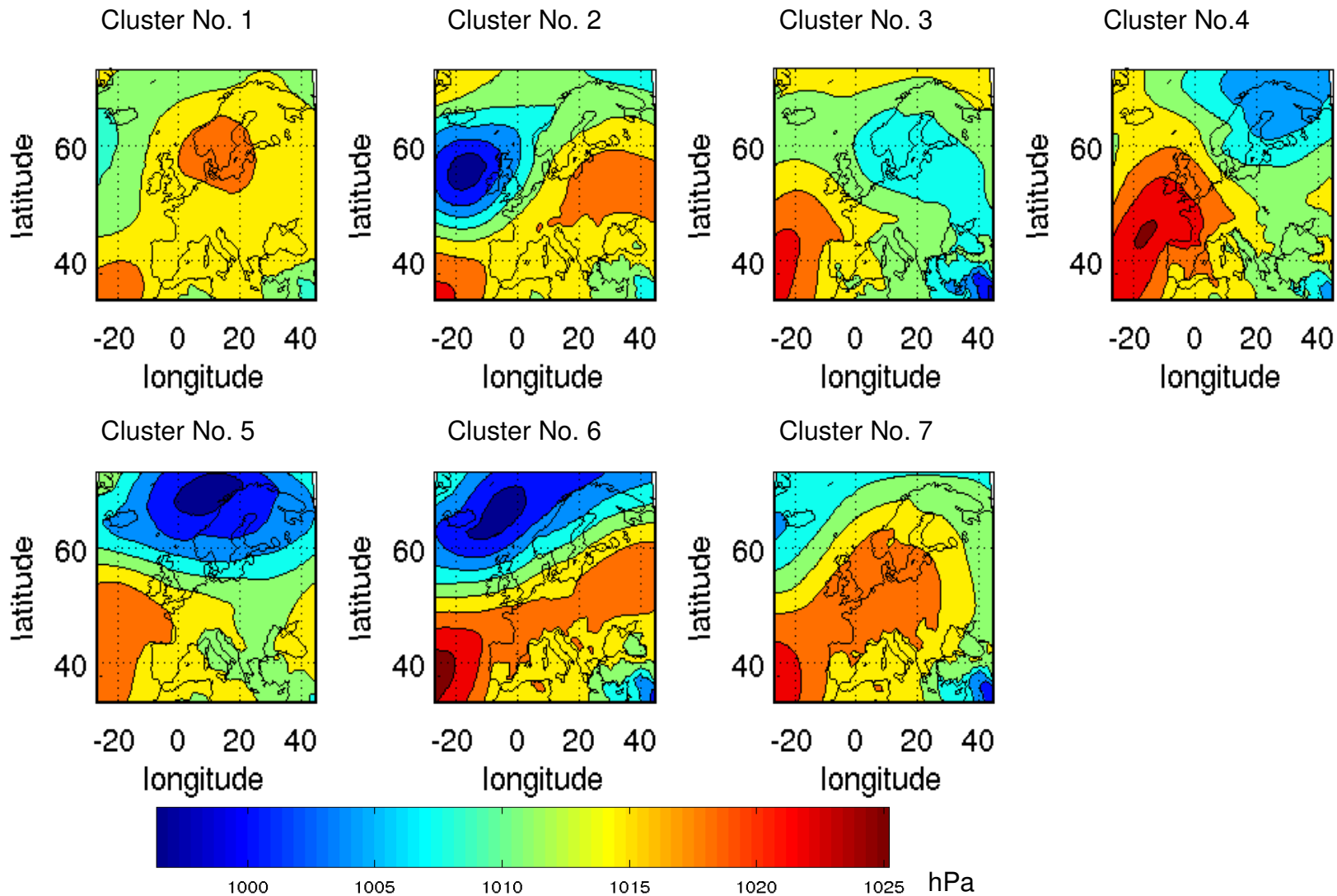
Summer - Cluster

500 hPa level January 2005 - April 2007



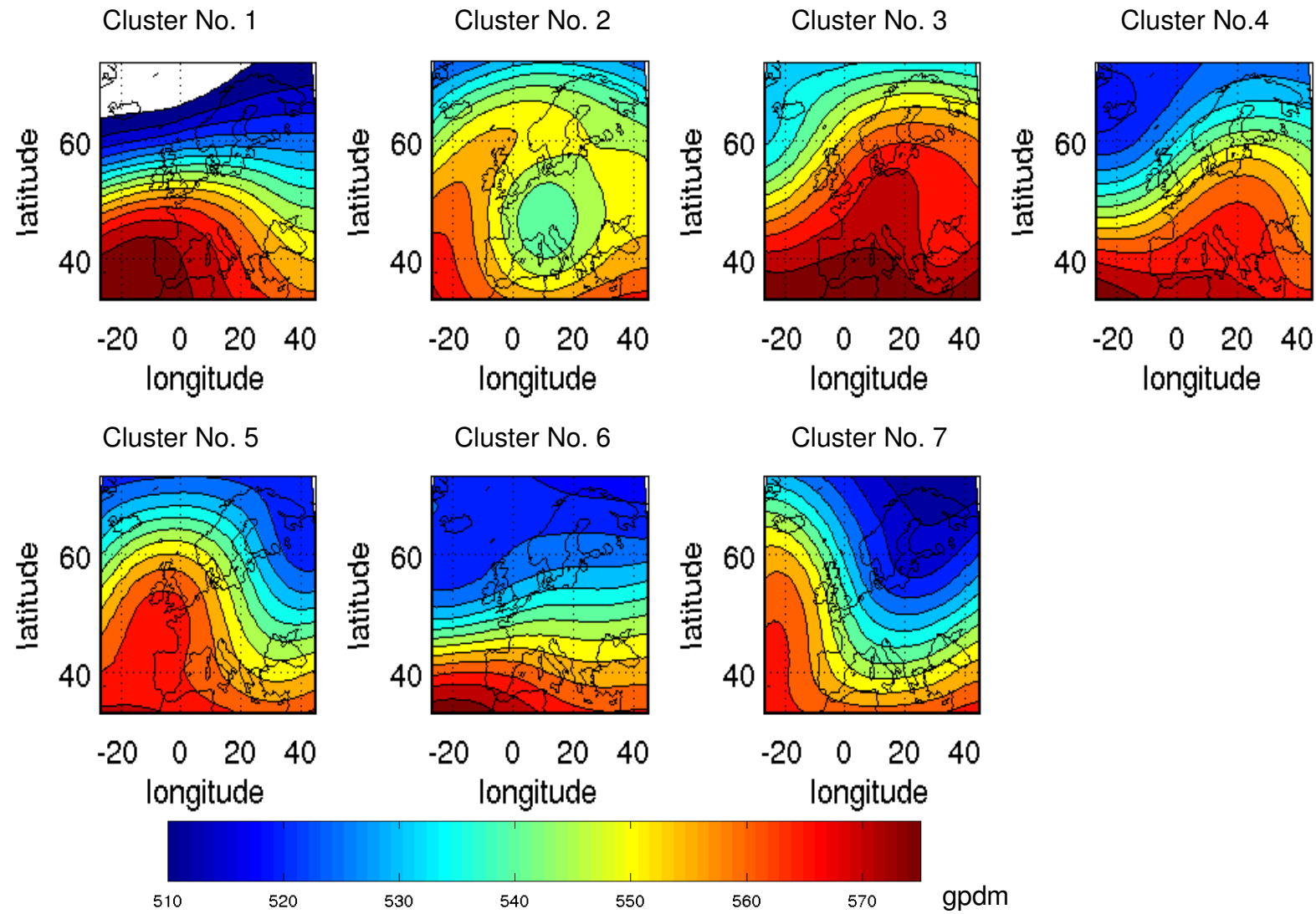
Summer - Cluster

sea level January 2005 - April 2007



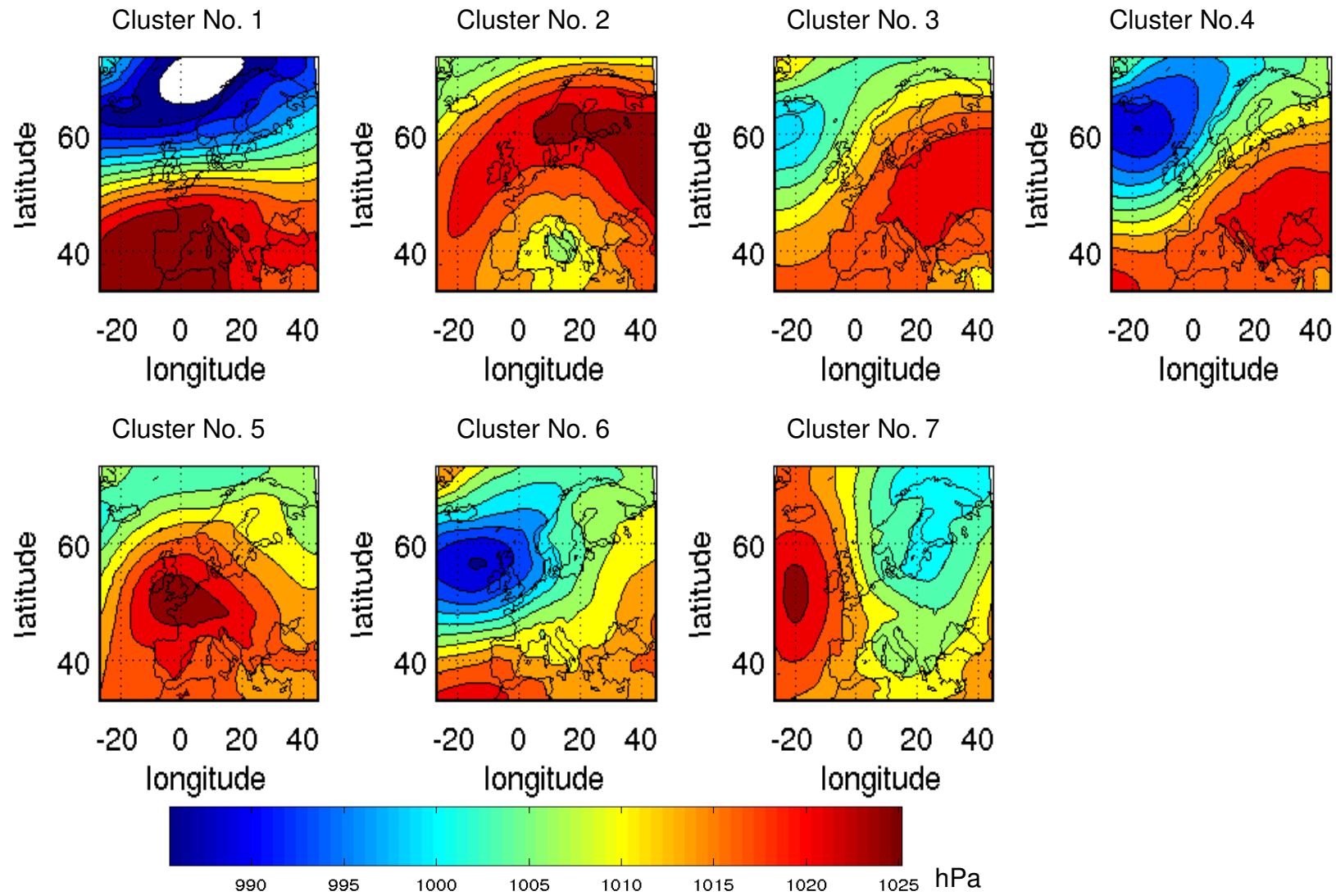
Winter - Cluster

500 hPa level January 2005 - April 2007



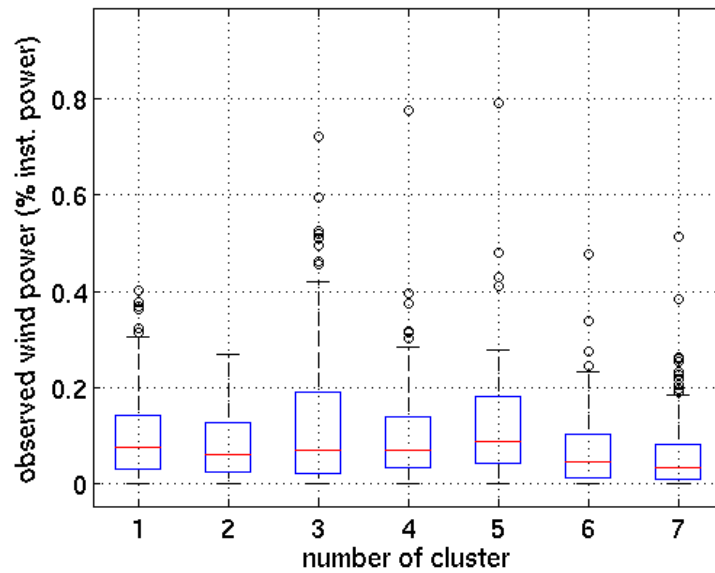
Winter - Cluster

sea level January 2005 - April 2007

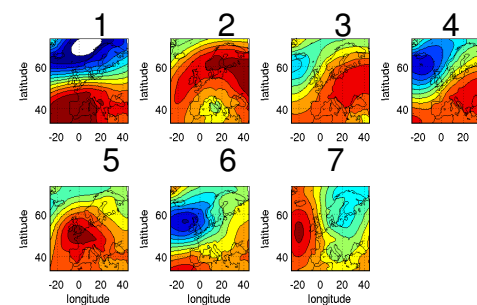
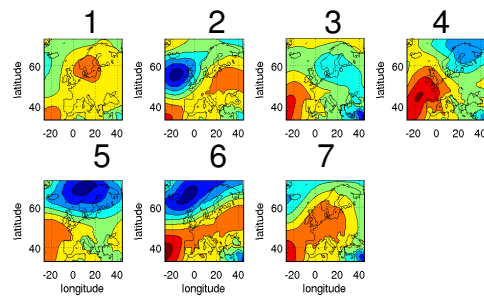
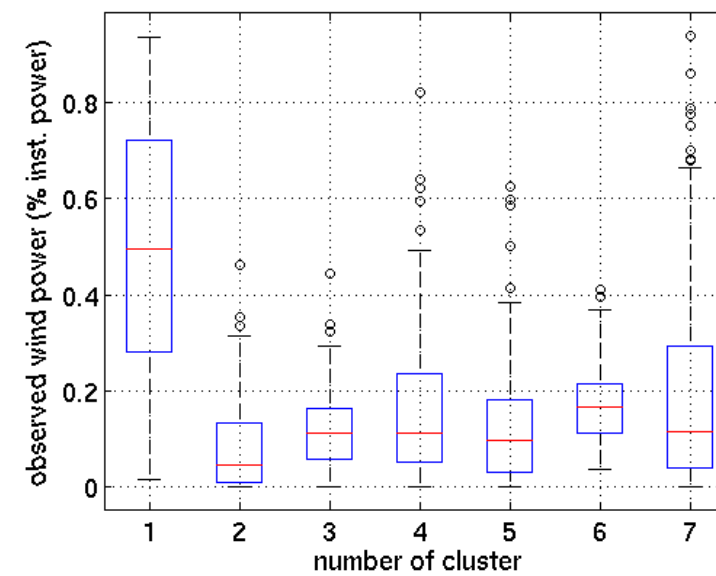


Observed wind power input for different clusters

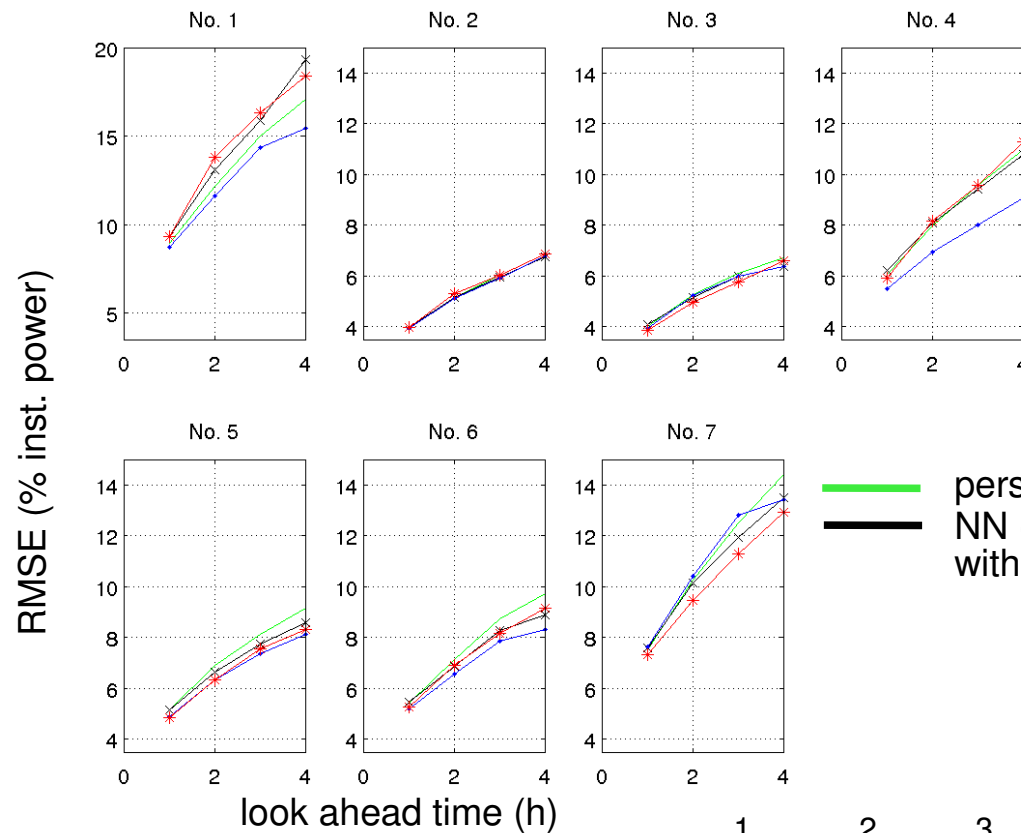
Summer



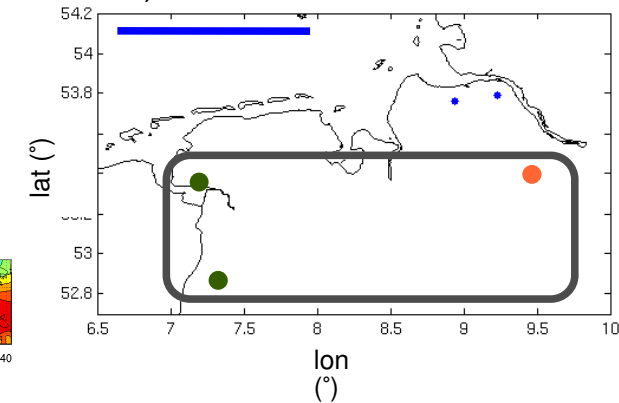
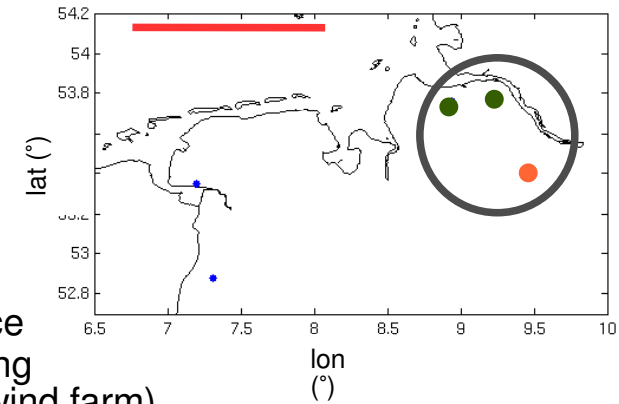
Winter



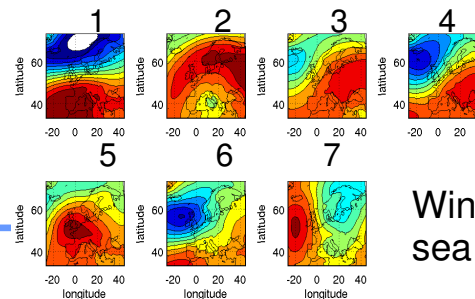
Forecast errors (RMSE) of wind power forecasts depending on clusters - winter



NN (training with three wind farms)



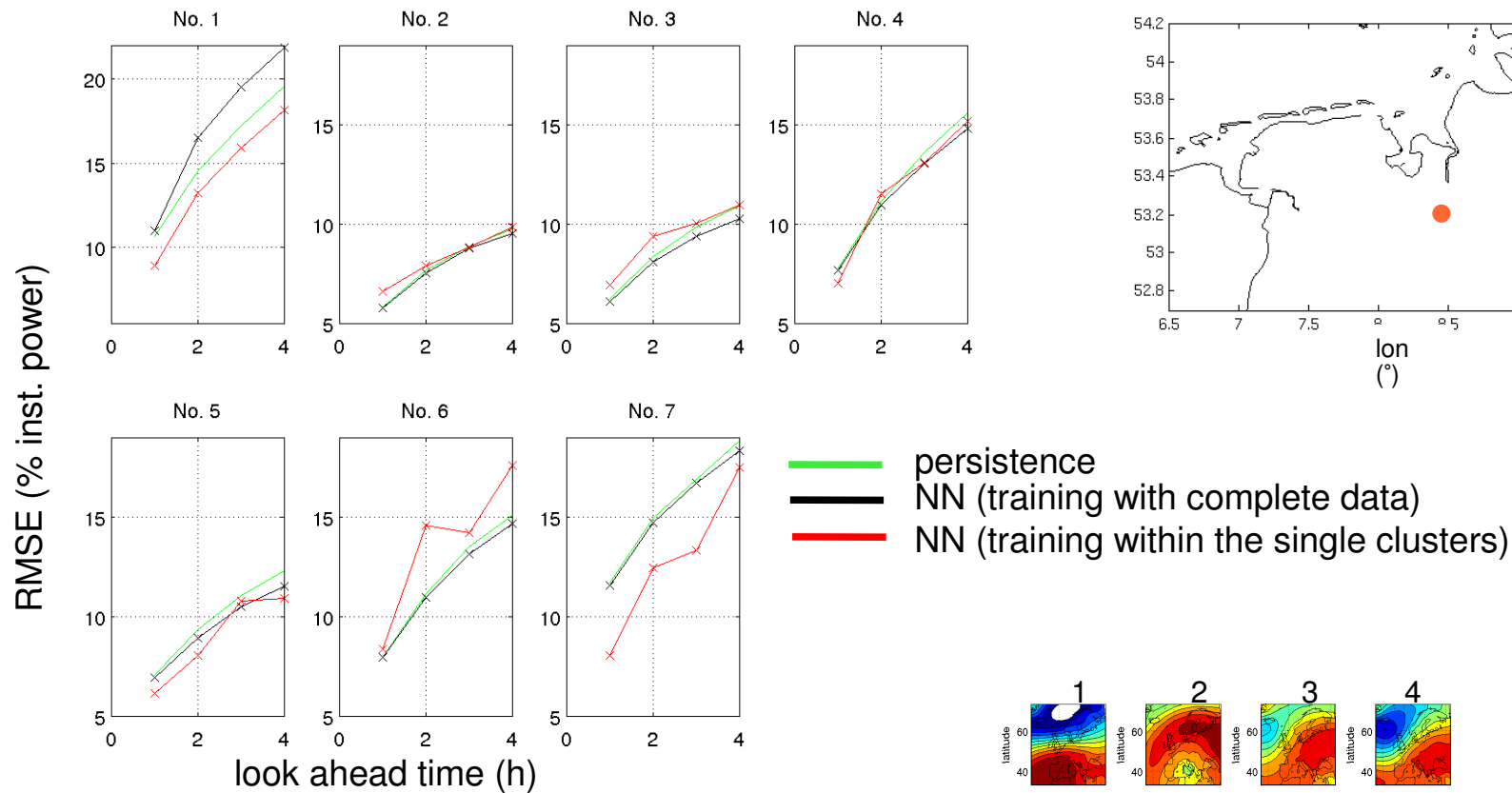
forecast:
one wind farm (●), 2005
training with data of 2004



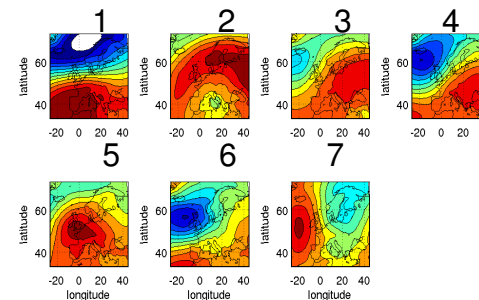
Winter cluster
sea level pressure

Training *within* clusters

Forecast errors (RMSE) - winter



forecast:
 one wind farm (●), 2006
 training with data of 2005,
 separately for each cluster



Conclusions

- near real time updates, require: near real time wind power data
- advantage: **no NWP data** necessary – very actual shortest term forecasts possible
- wind power input and forecast errors depend on **weather situation** (clusters)
- for some clusters **improvements** are possible (as shown):
- consideration of geographical distribution of the wind farms
- training differentiation by clusters

Perspectives

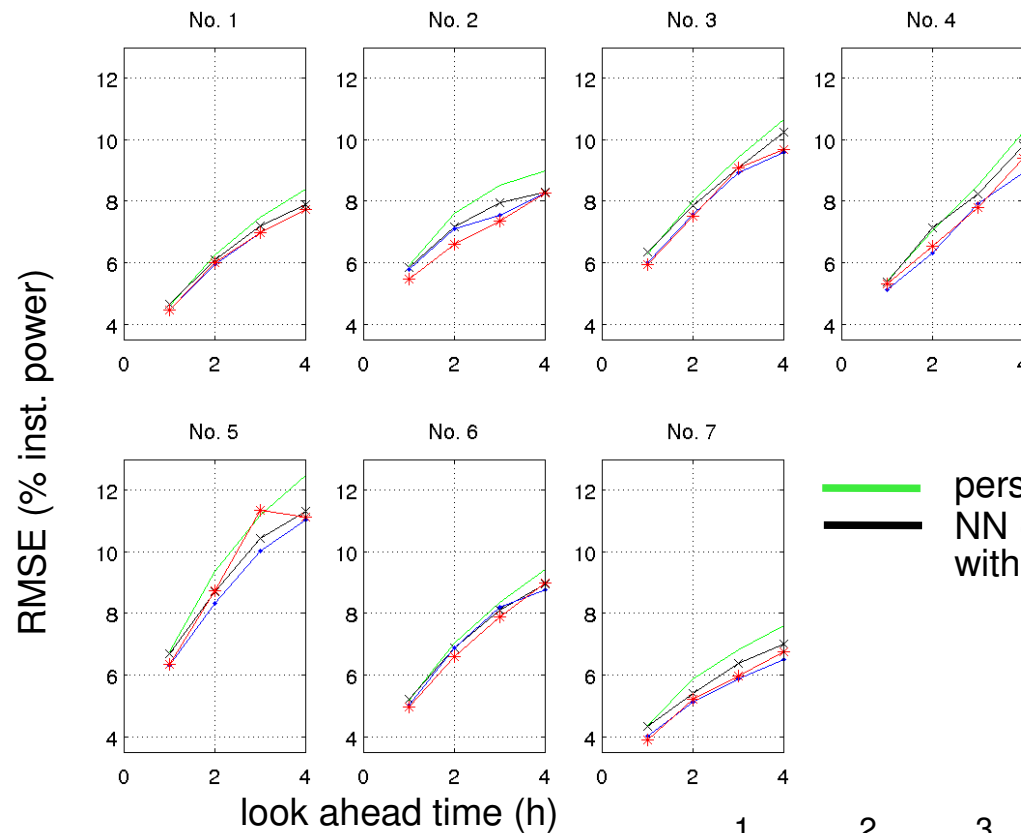
- larger data set including **more wind farms**
- more sophisticated methods to capture **spatial patterns**
- apply **different methods**: Neural Networks, autoregressive models
- combine with model using NWP

Thank you for your attention.

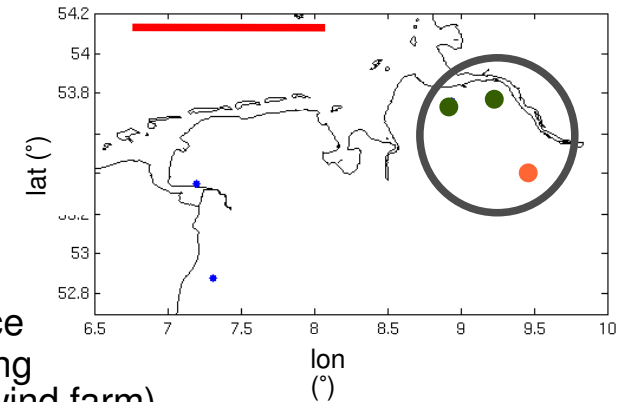
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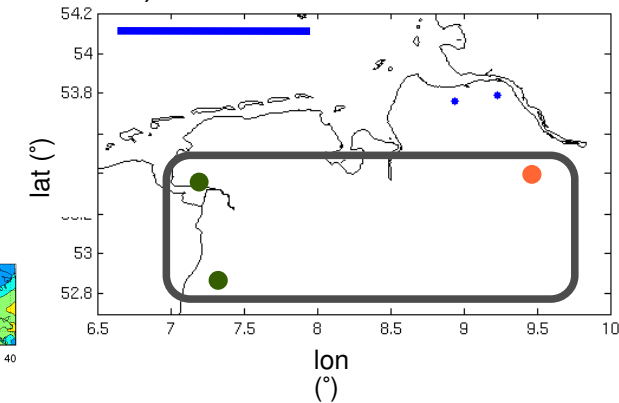
Forecast errors (RMSE) of wind power forecasts depending on clusters - summer



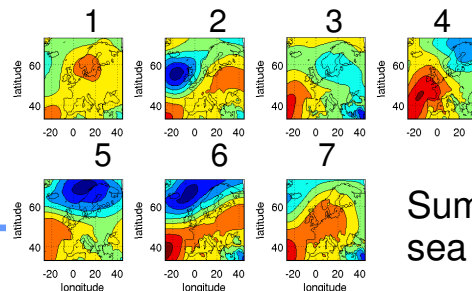
NN (training with three wind farms)



— persistence
— NN (training with one wind farm)

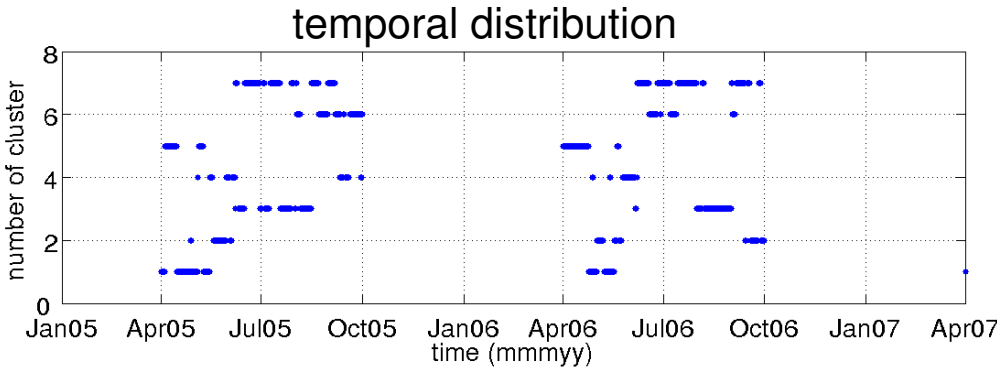


forecast:
one wind farm (●), 2005
training with data of 2004

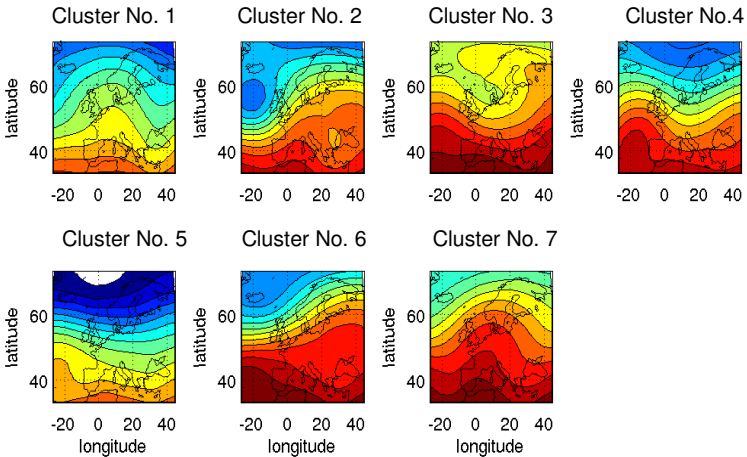
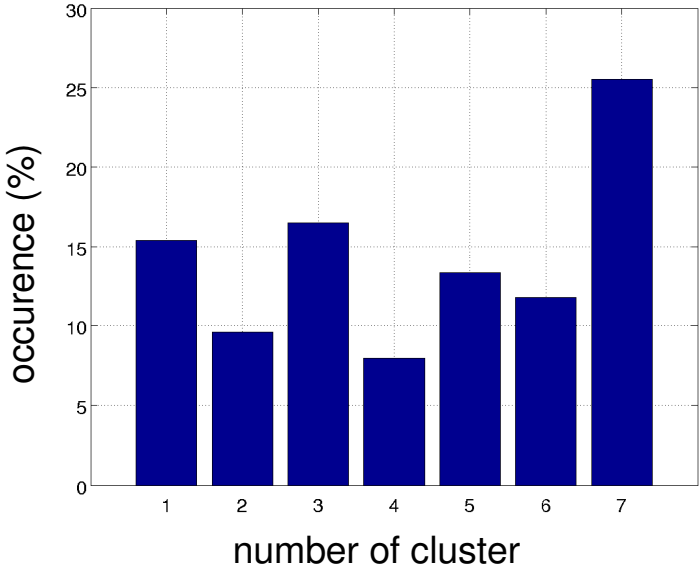


Summer cluster
sea level pressure

Summer – Cluster

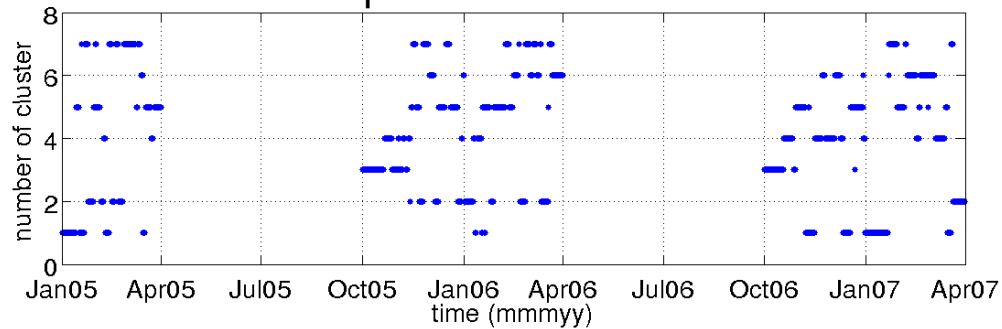


occurrence of each cluster (%)



Winter – Cluster

temporal distribution



occurrence of each cluster (%)

