

Local and regional PV power forecasting based on PV measurements, satellite data and numerical weather predictions

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University of Oldenburg



Outline

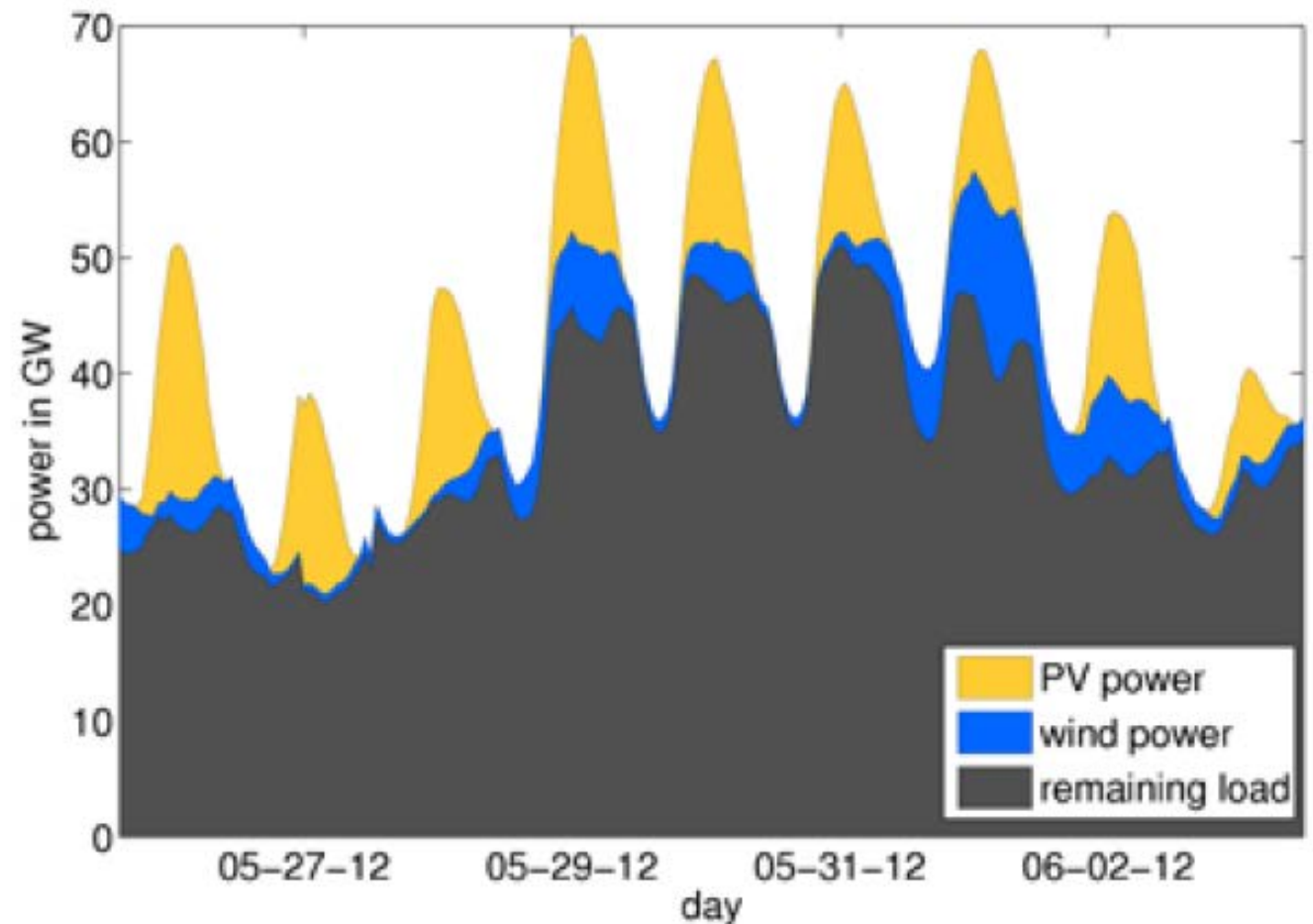
- ▶ background: grid integration of PV power in Germany
- ▶ overview on PV power prediction system
- ▶ evaluation:
 - ▶ different data and models for different forecast horizons
 - ▶ combination of different models
- ▶ summary and outlook

Contribution of photovoltaic (PV) systems to electricity supply in Germany

- ▶ installed PV power (end 2015):

39.5 GW_{peak}

- ▶ up to **50 %** of electricity demand from PV

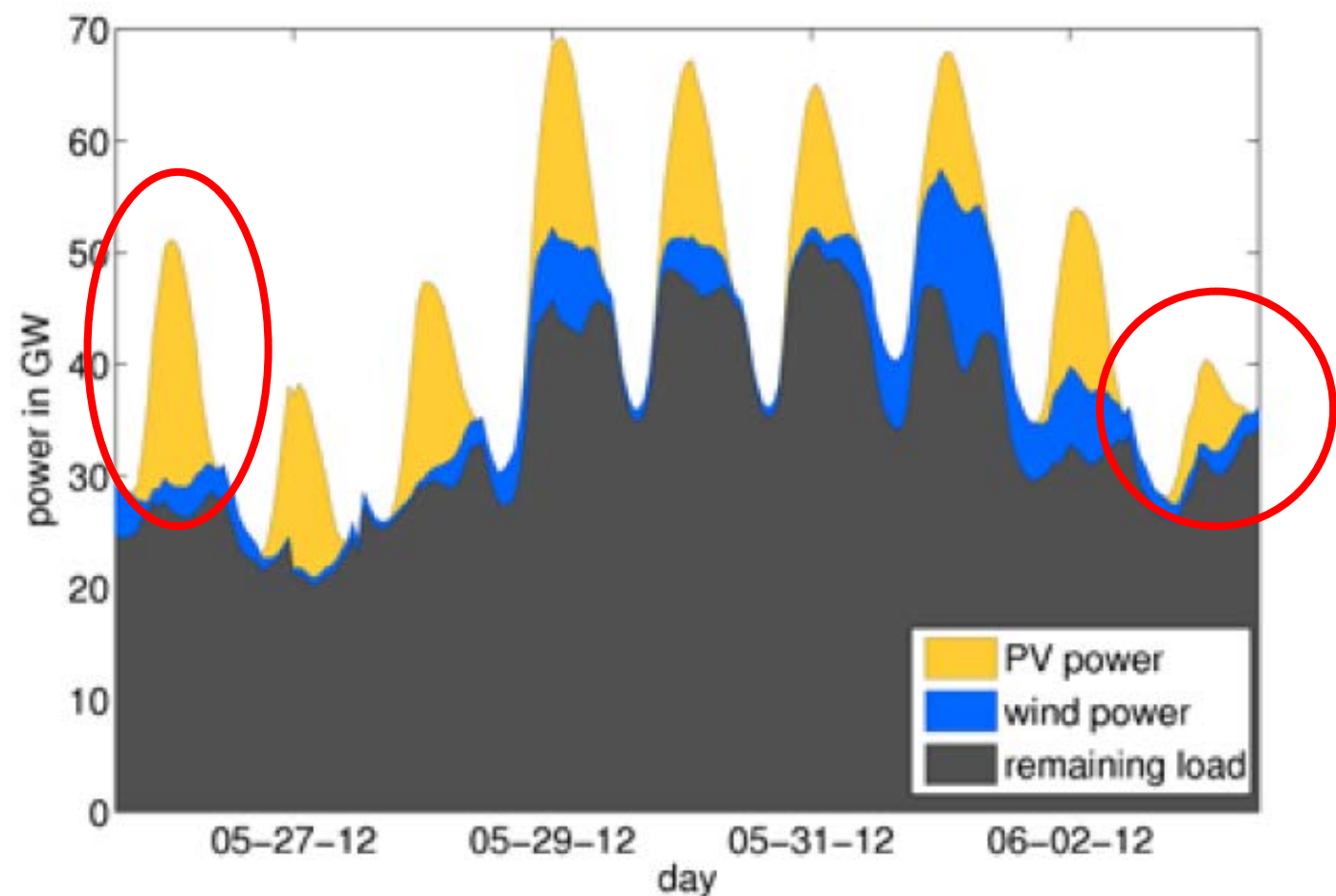


Contribution of photovoltaic (PV) systems to electricity supply in Germany

- ▶ installed PV power (end 2015):

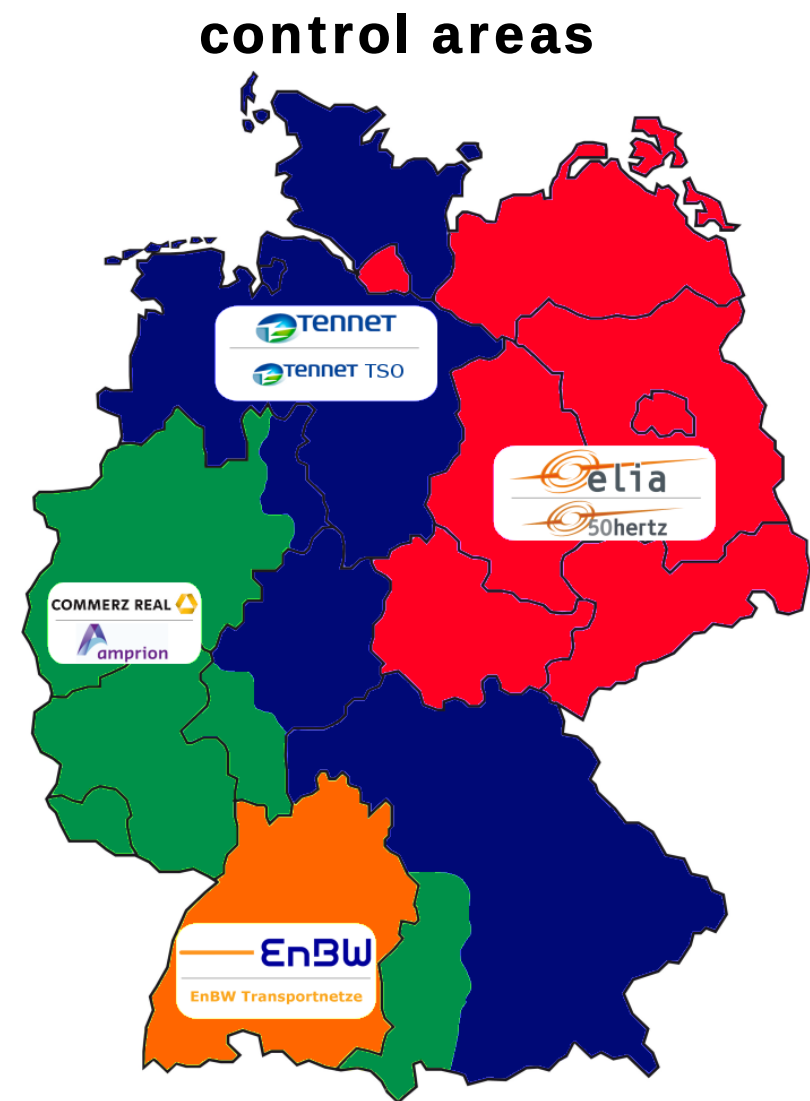
39.5 GW_{peak}

- ▶ up to **50 %** of electricity demand from PV
- ▶ strong variability of solar power



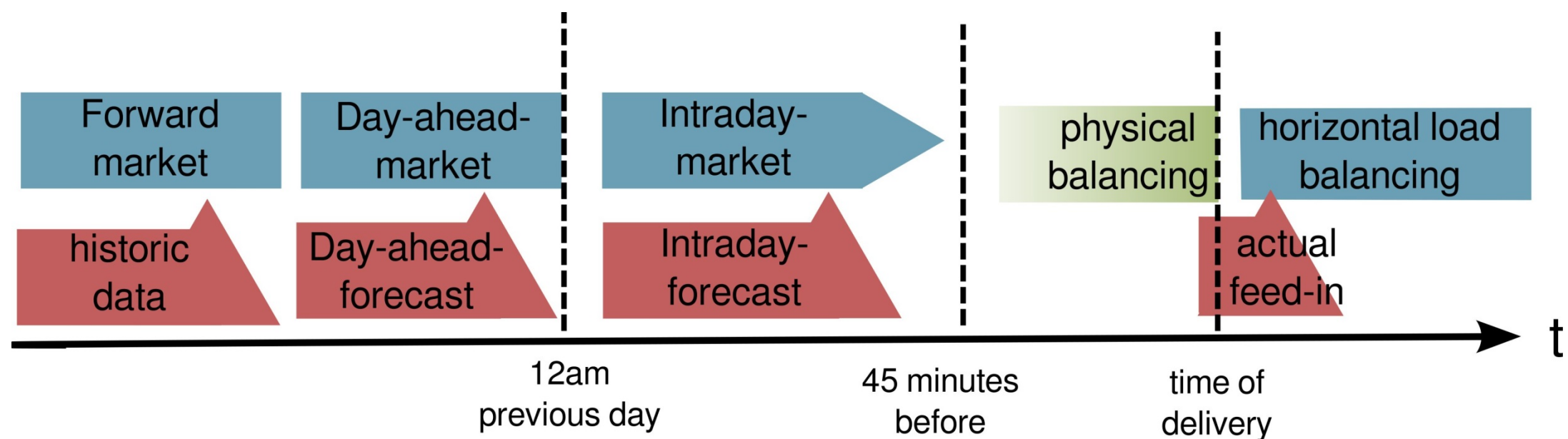
Grid integration of PV Power: marketing at the European Energy Exchange

- ▶ by Transmission System Operators
 ➡ regional forecasts
- ▶ direct marketing
 ➡ local forecasts



Grid integration of PV Power: marketing at the European Energy Exchange

- ▶ by Transmission System Operators
→ regional forecasts
- ▶ direct marketing
→ local forecasts
- ▶ forecast horizons



Grid integration of PV Power: marketing at the European Energy Exchange

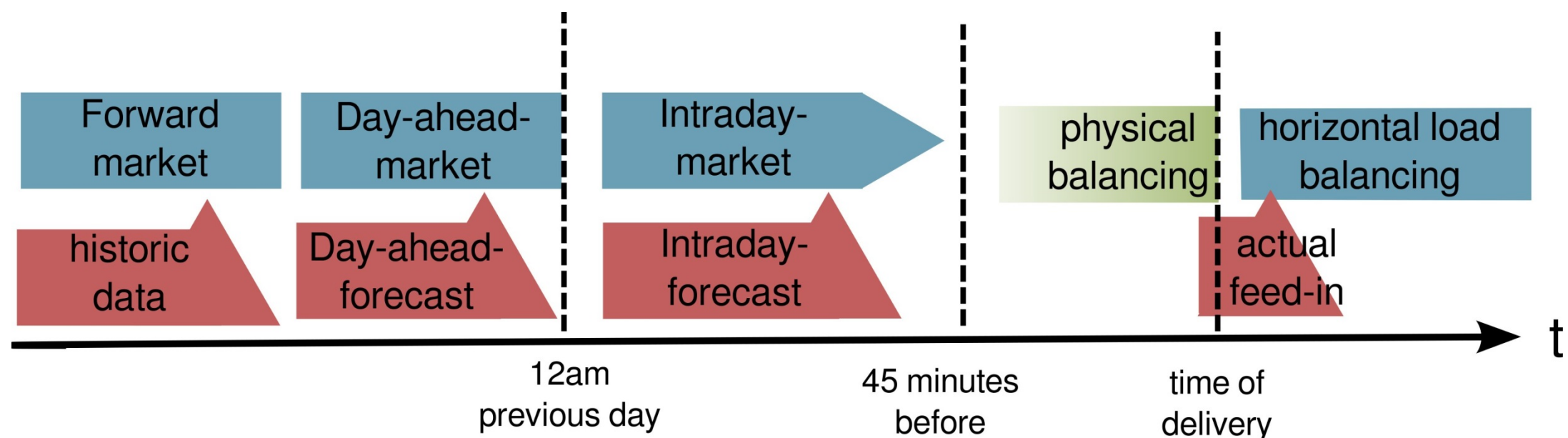
- ▶ by Transmission System Operators
→ regional forecasts

- ▶ direct marketing
→ local forecasts

- ▶ forecast horizons

forecast errors *high price*

*need for
balancing power*

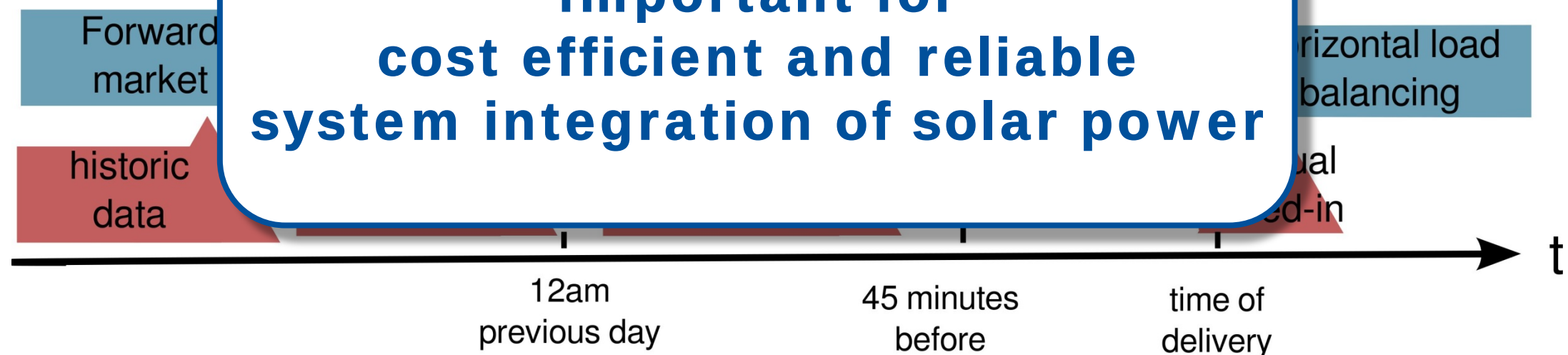


Grid integration of PV Power: marketing at the European Energy Exchange

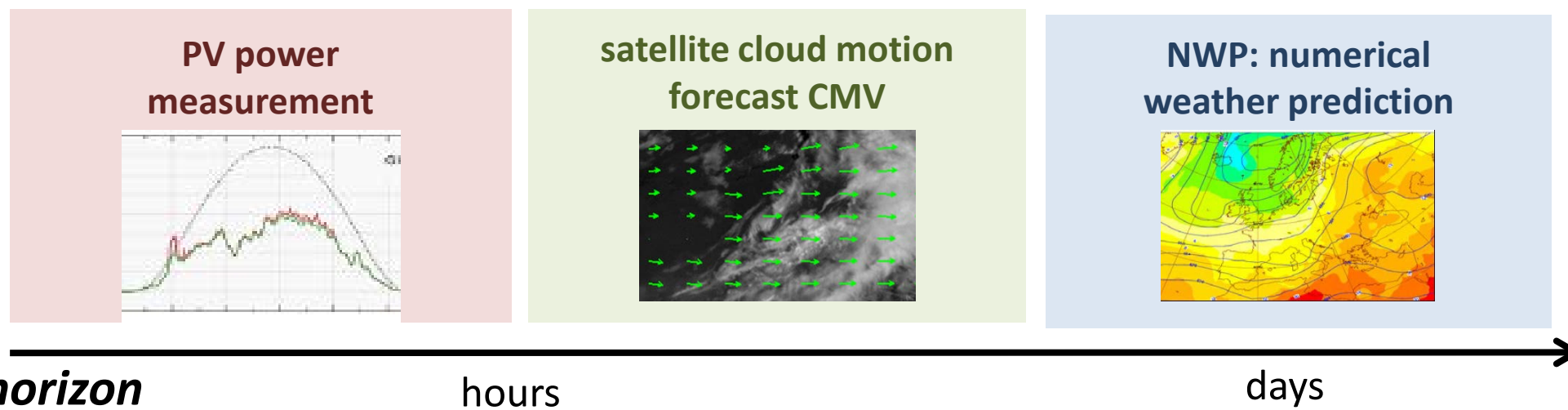
- ▶ by Transmission System Operators
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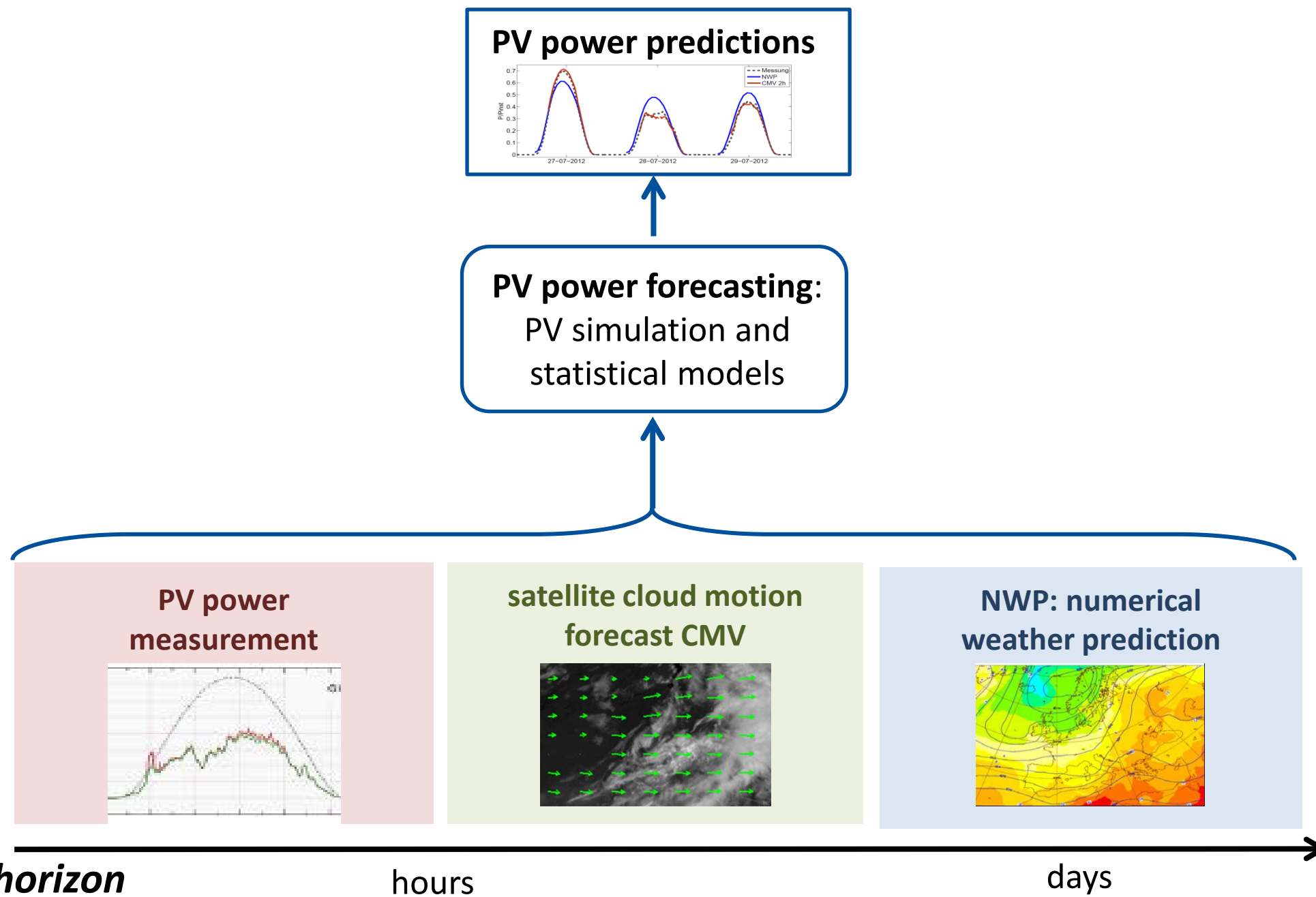
- ▶ forecast horizons
forecast



Overview of forecasting scheme



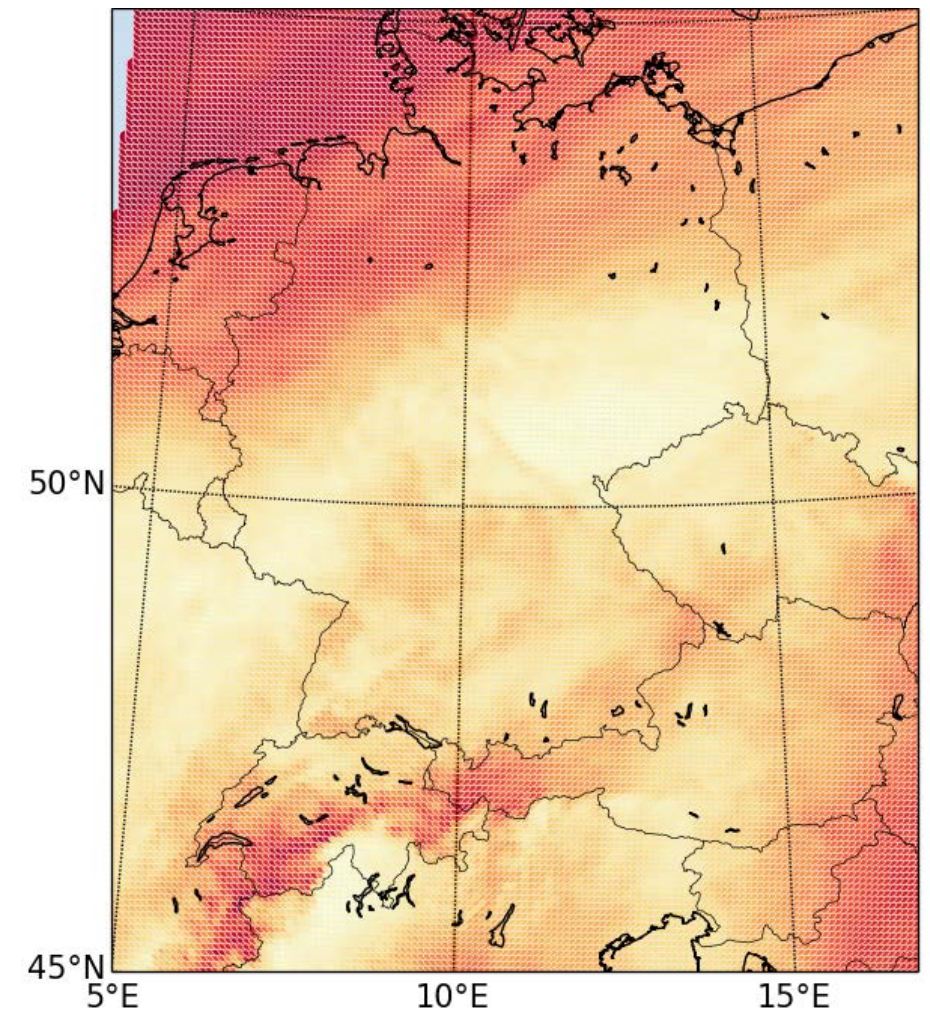
Overview of forecasting scheme



Numerical weather predictions

- ▶ global model forecast (IFS) of the European Centre for Medium-Range Weather Forecasts (ECWMF)
- ▶ regional model forecasts (COSMO EU) of the German Meteorological service (DWD)

COSMO EU, dir. irradiance
2104-05-02, 12:00



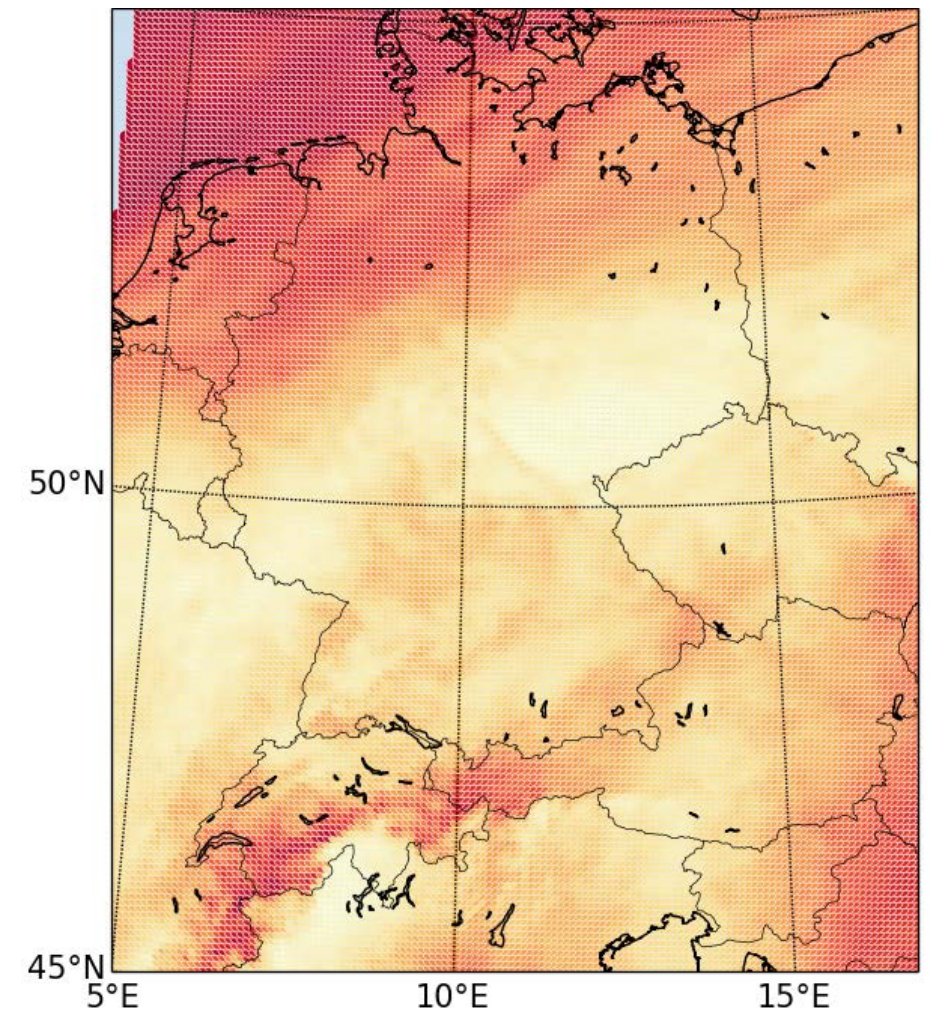
Numerical weather predictions

- ▶ global model forecast (IFS) of the European Centre for Medium-Range Weather Forecasts (ECWMF)
- ▶ regional model forecasts (COSMO EU) of the German Meteorological service (DWD)

Post processing:

- ▶ bias correction and combination with linear regression

COSMO EU, dir. irradiance
2104-05-02, 12:00



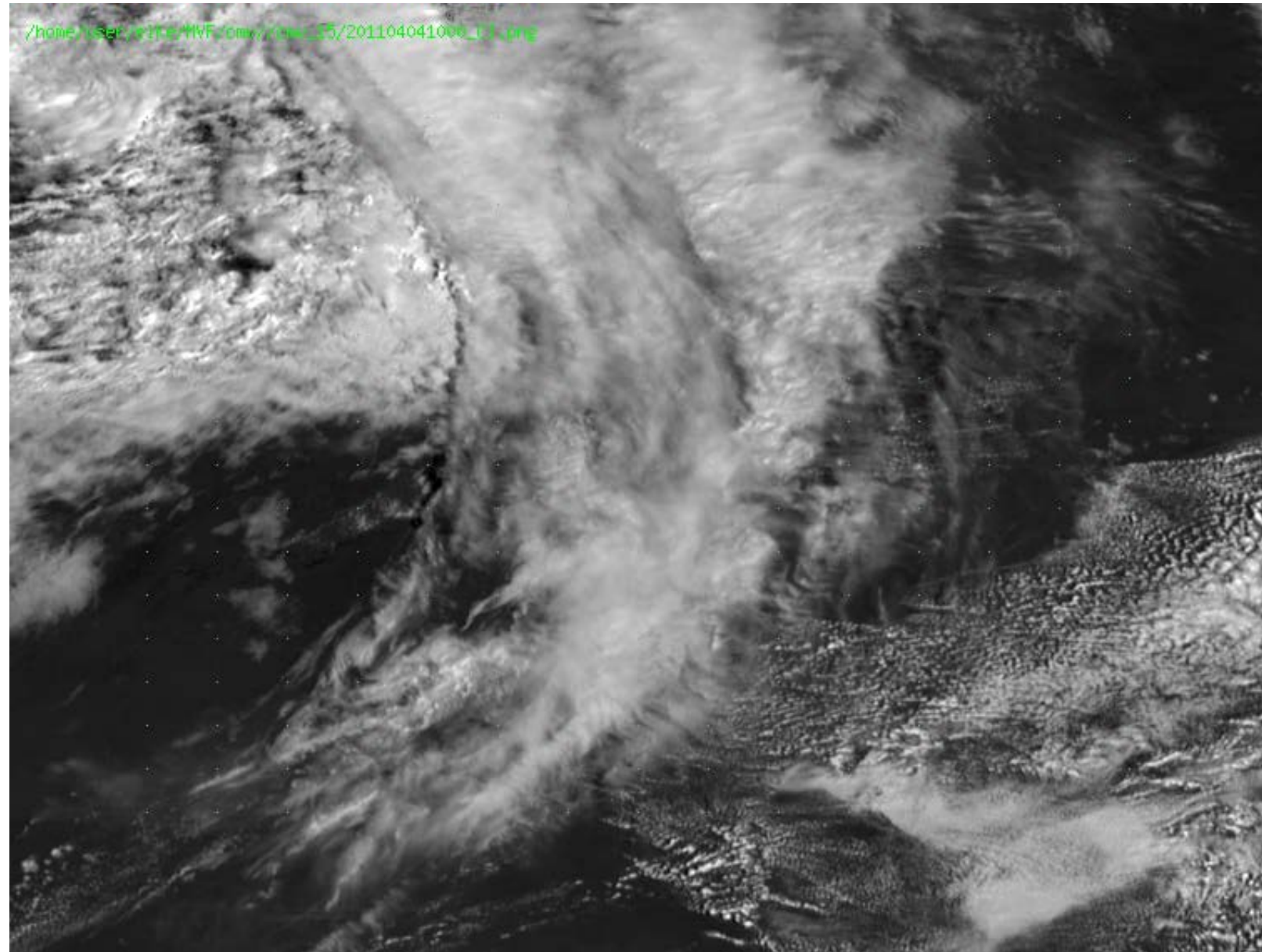
Irradiance prediction based on satellite data

- ▶ cloud index from
Meteosat images with
Heliosat method*

resolution in Germany

- ▶ 1.2km x 2.2 km
- ▶ 15 minutes

Meteosat Second Generation (high resolution visible range)

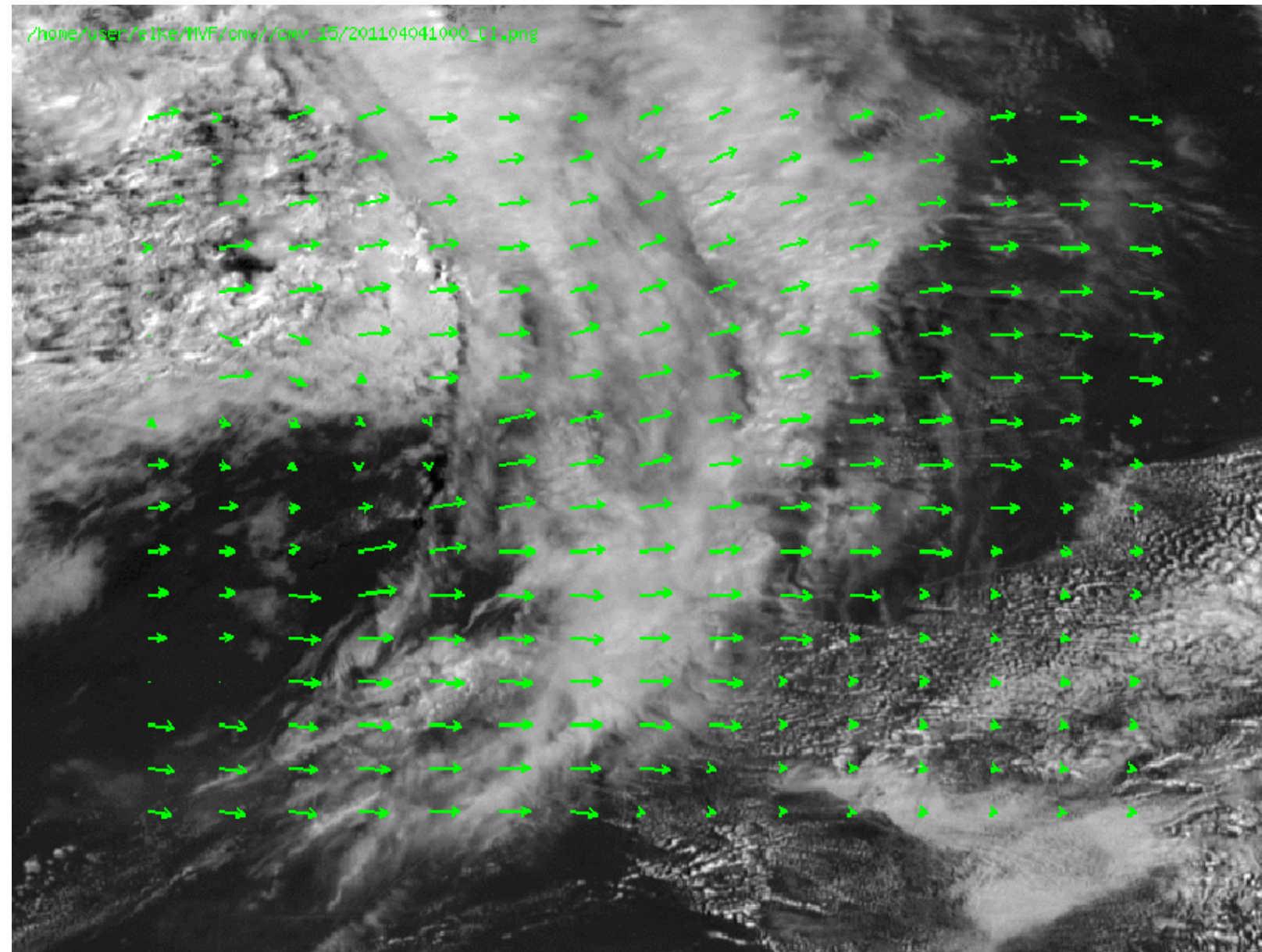


* Hammer et al 2003

Irradiance prediction based on satellite data

- ▶ cloud index from
Meteosat images with
Heliosat method
- ▶ cloud motion vectors by
identification of matching
cloud structures in
consecutive images
- ▶ extrapolation of
cloud motion to predict
future cloud index

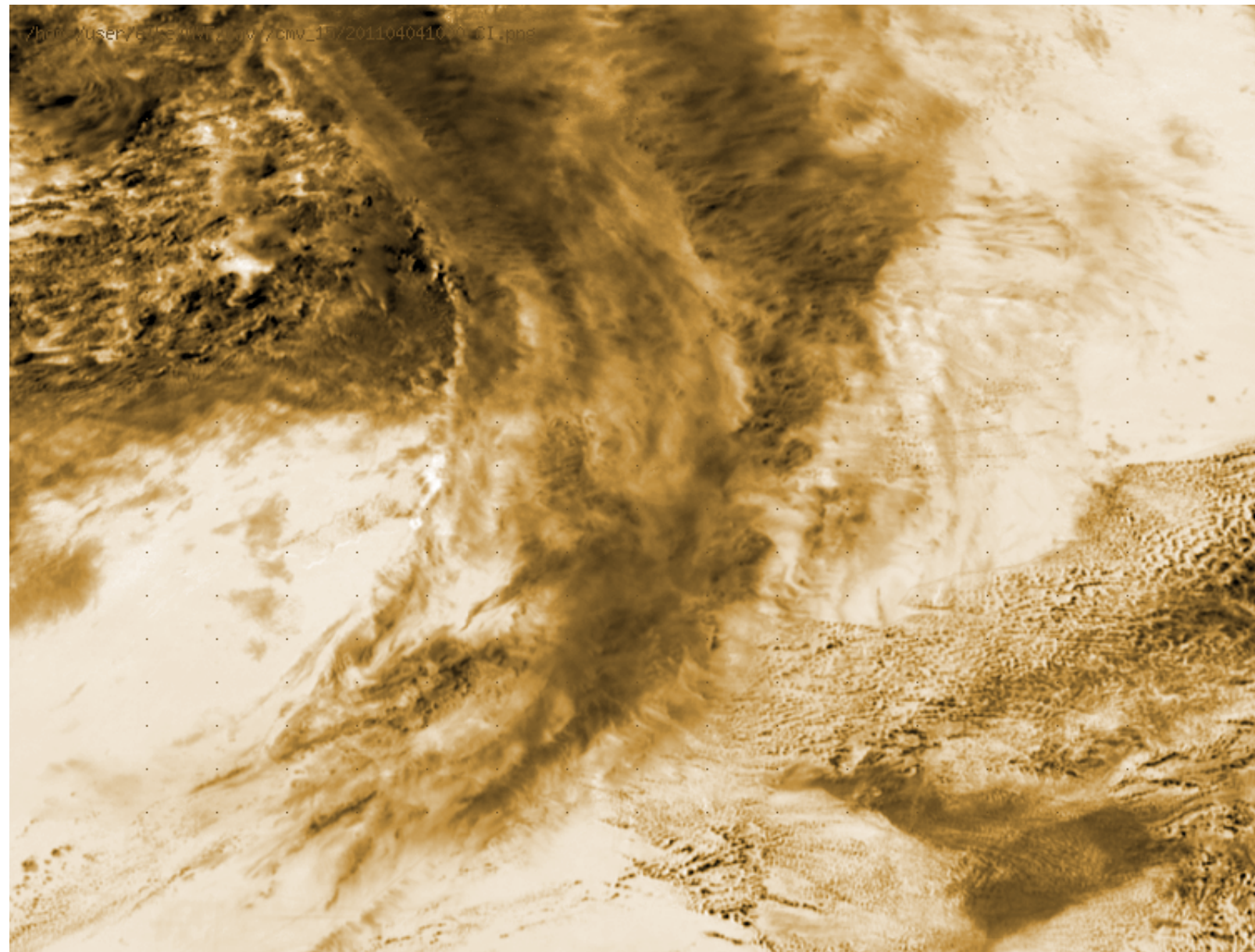
Meteosat Second Generation (high resolution visible range)



Irradiance prediction based on satellite data

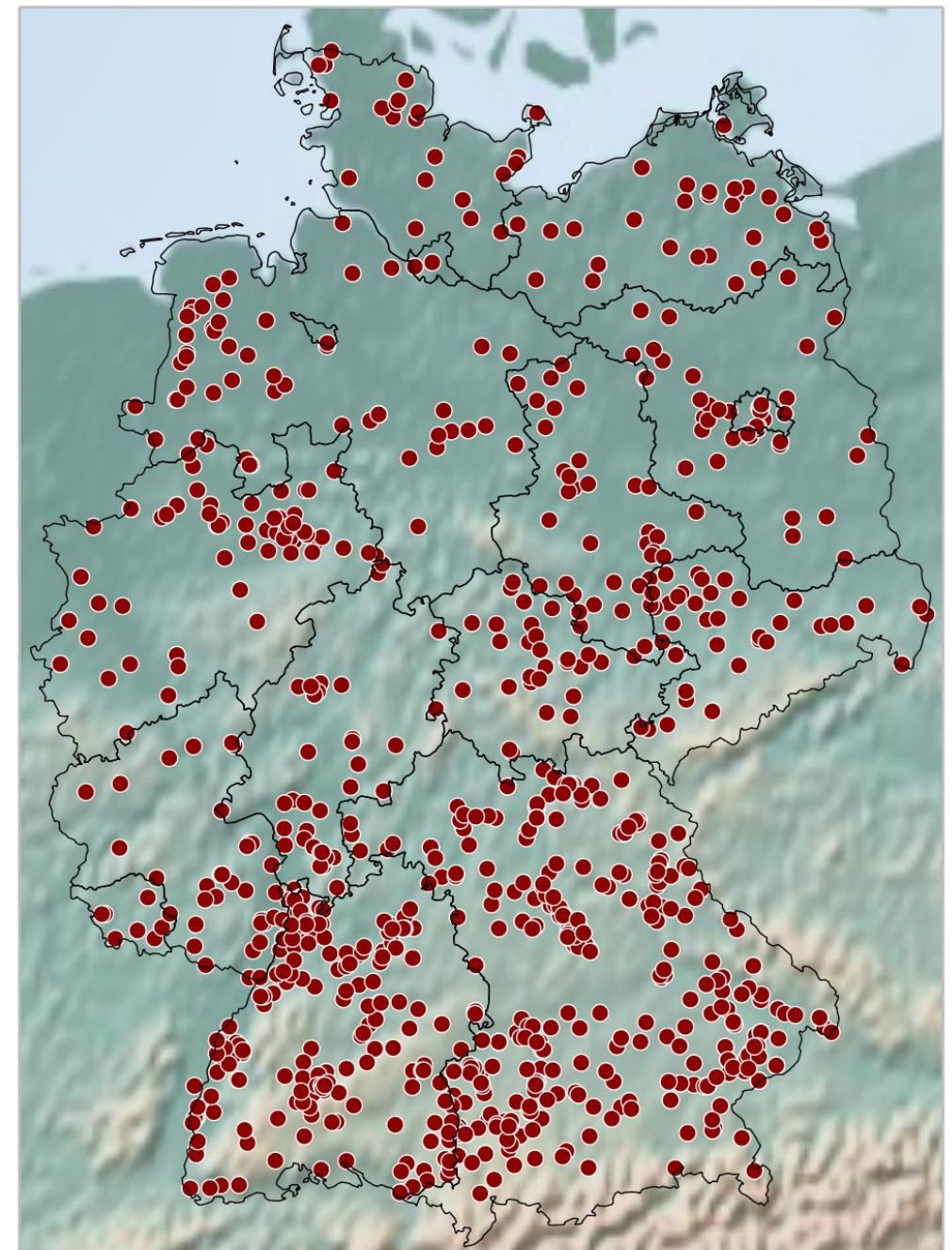
- ▶ cloud index from Meteosat images with Heliosat method
- ▶ cloud motion vectors by identification of matching cloud structures in consecutive images
- ▶ extrapolation of cloud motion to predict future cloud index
- ▶ irradiance from predicted cloud index images with Heliosat method

satellite derived irradiance maps



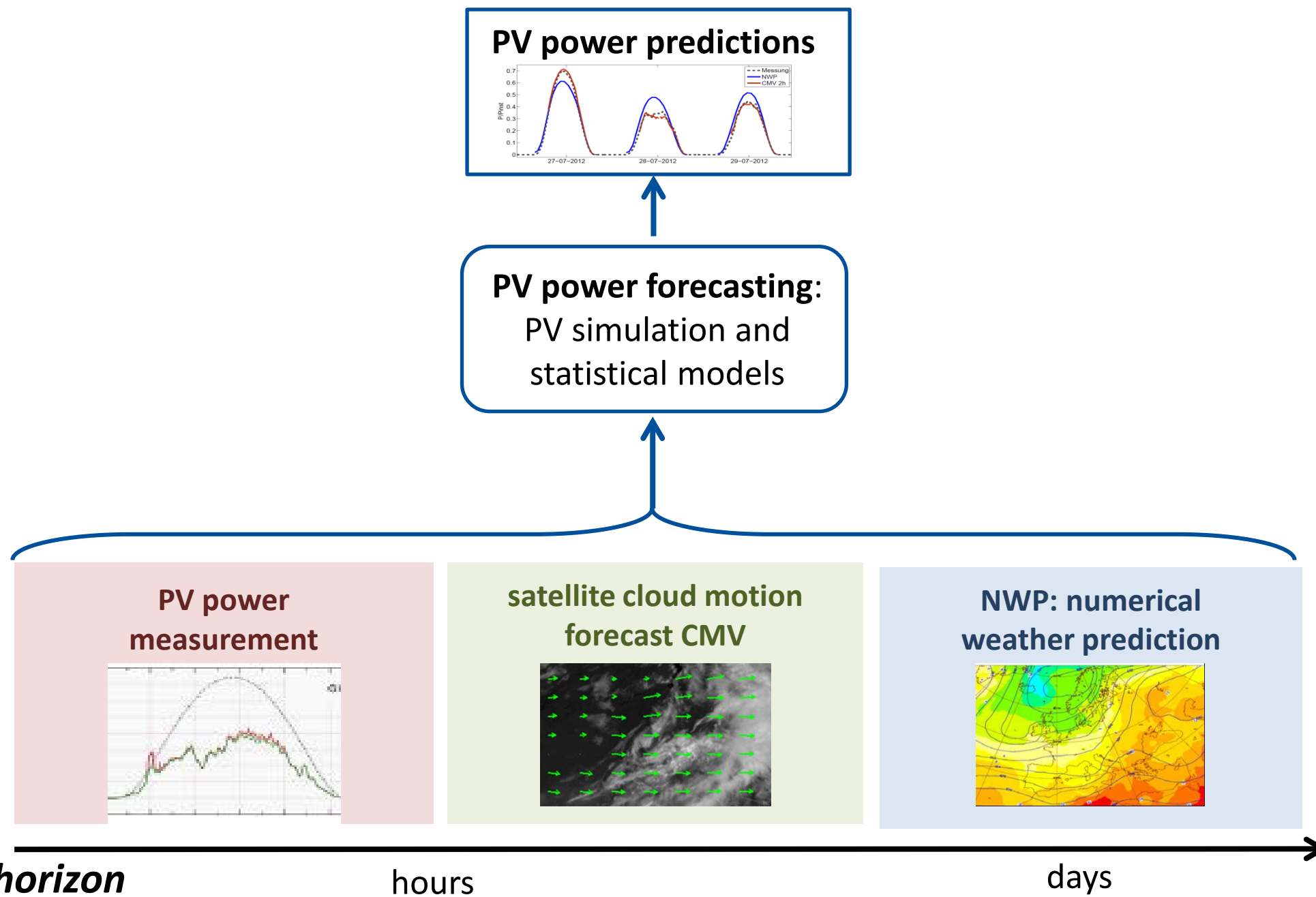
Measurement data

- ▶ March- November 2013
- ▶ 15 minute values
- ▶ 921 PV systems¹⁾ in Germany
- ▶ information on PV system tilt and orientation



¹⁾Monitoring data base of Meteocontrol GmbH

Overview of forecasting scheme



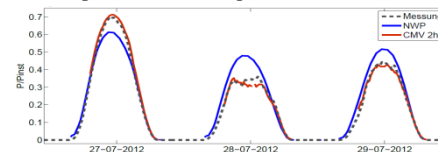
Overview of forecasting scheme

persistence:

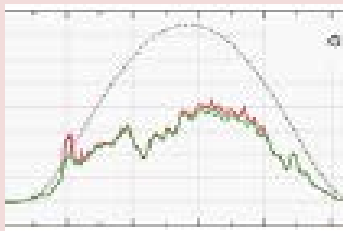
constant ratio of
measured PV power P_{meas}
to clear sky PV power P_{clear}

$$P_{\text{pers}}(t) = \frac{P_{\text{meas}}(t-\Delta t)}{P_{\text{clear}}(t-\Delta t)} P_{\text{clear}}(t)$$

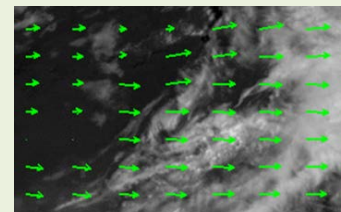
PV power predictions



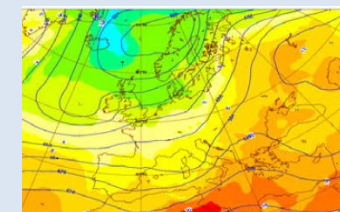
PV power measurement



satellite cloud motion forecast CMV



NWP: numerical weather prediction

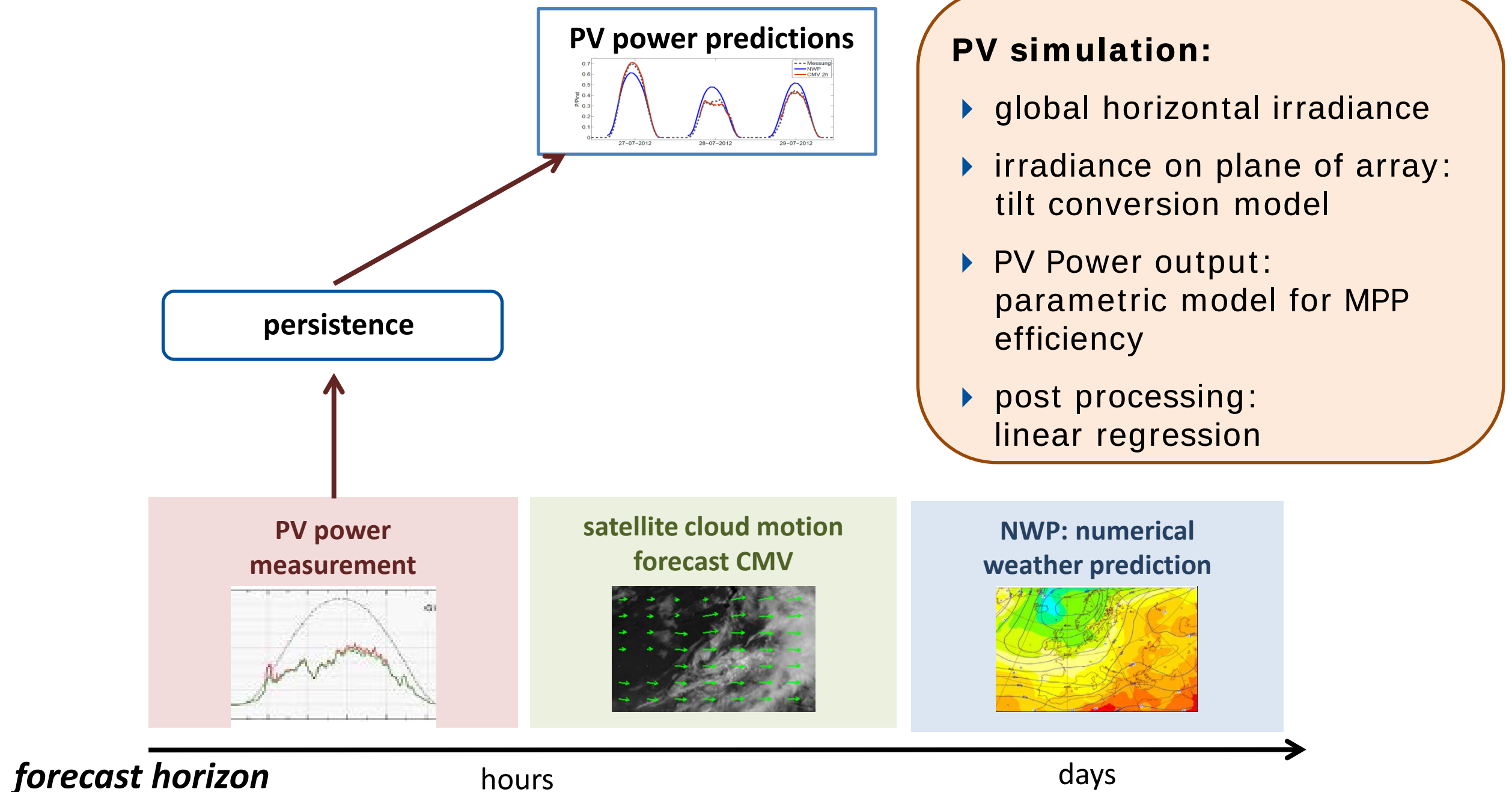


forecast horizon

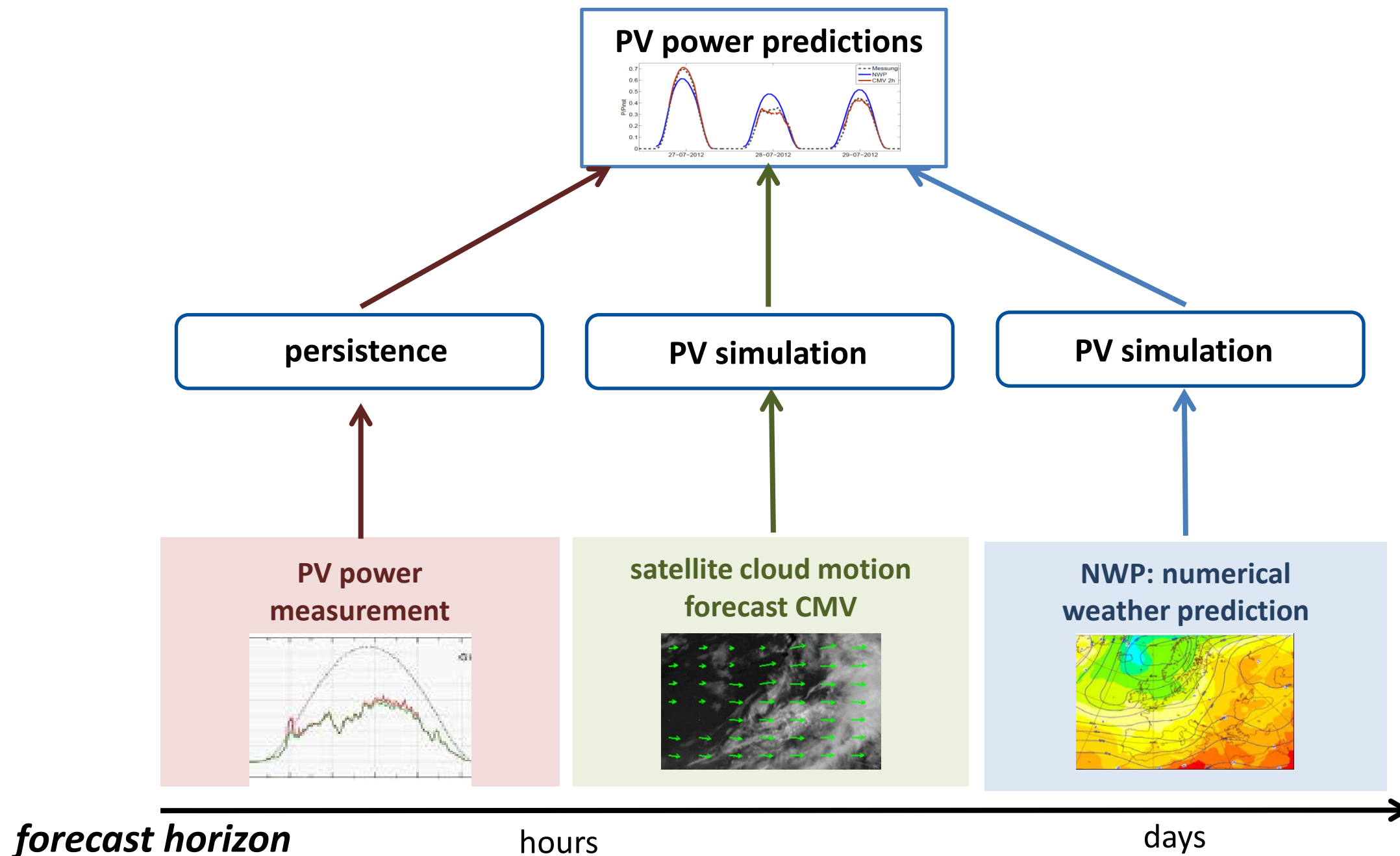
hours

days

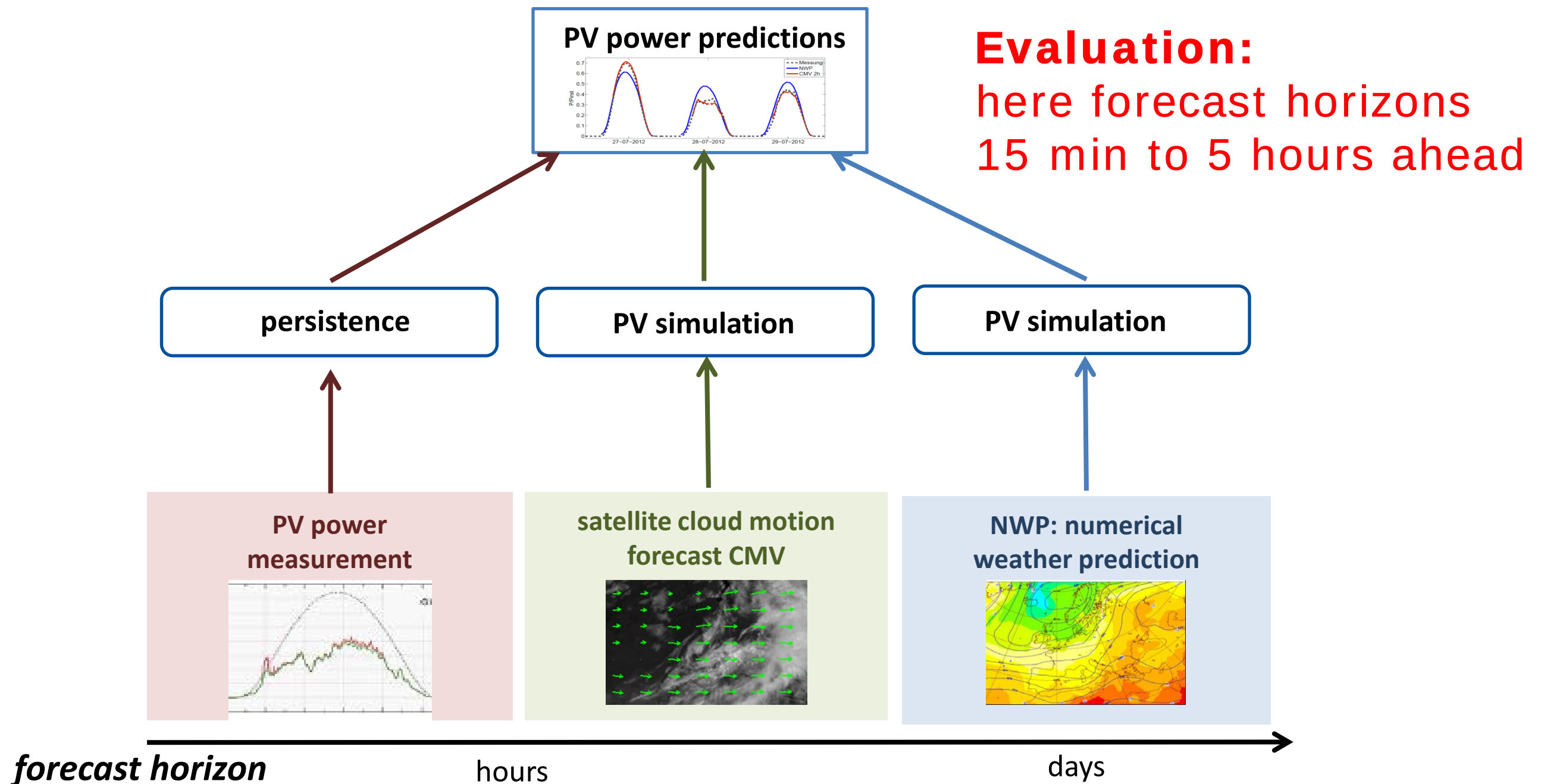
Different input data and models



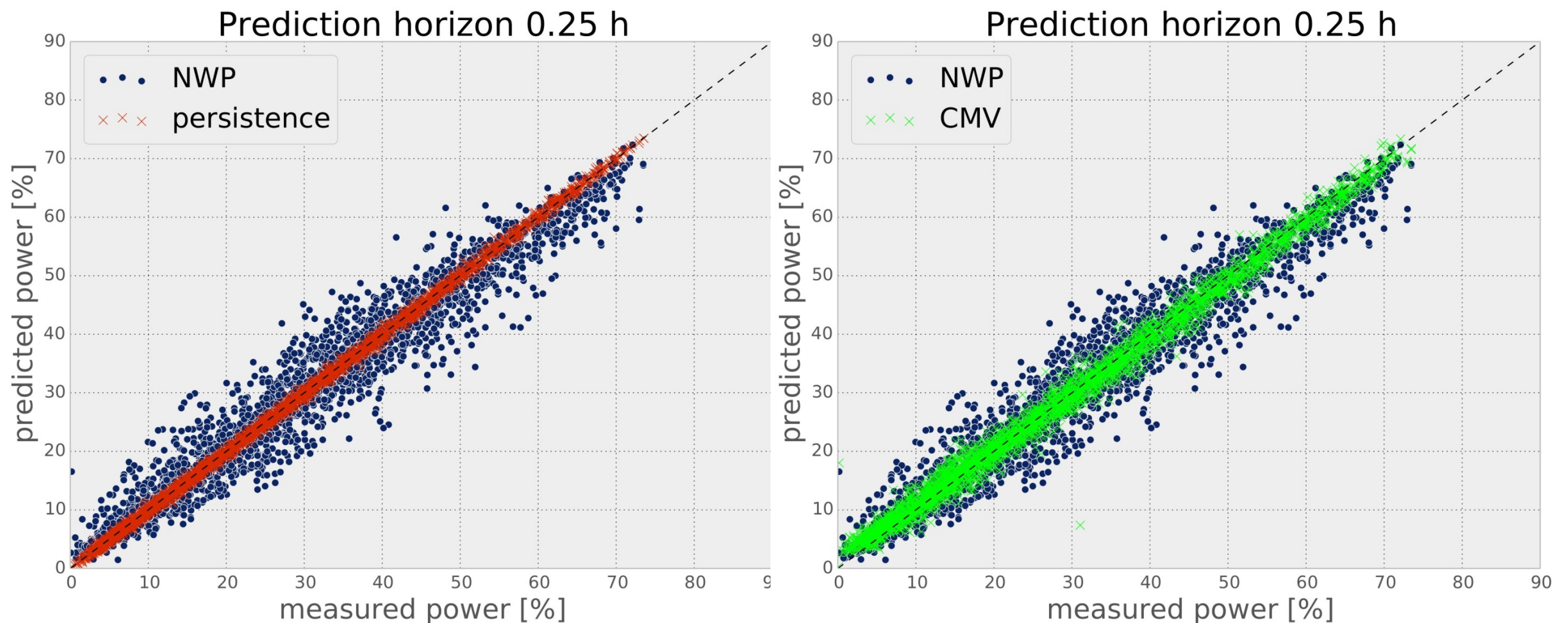
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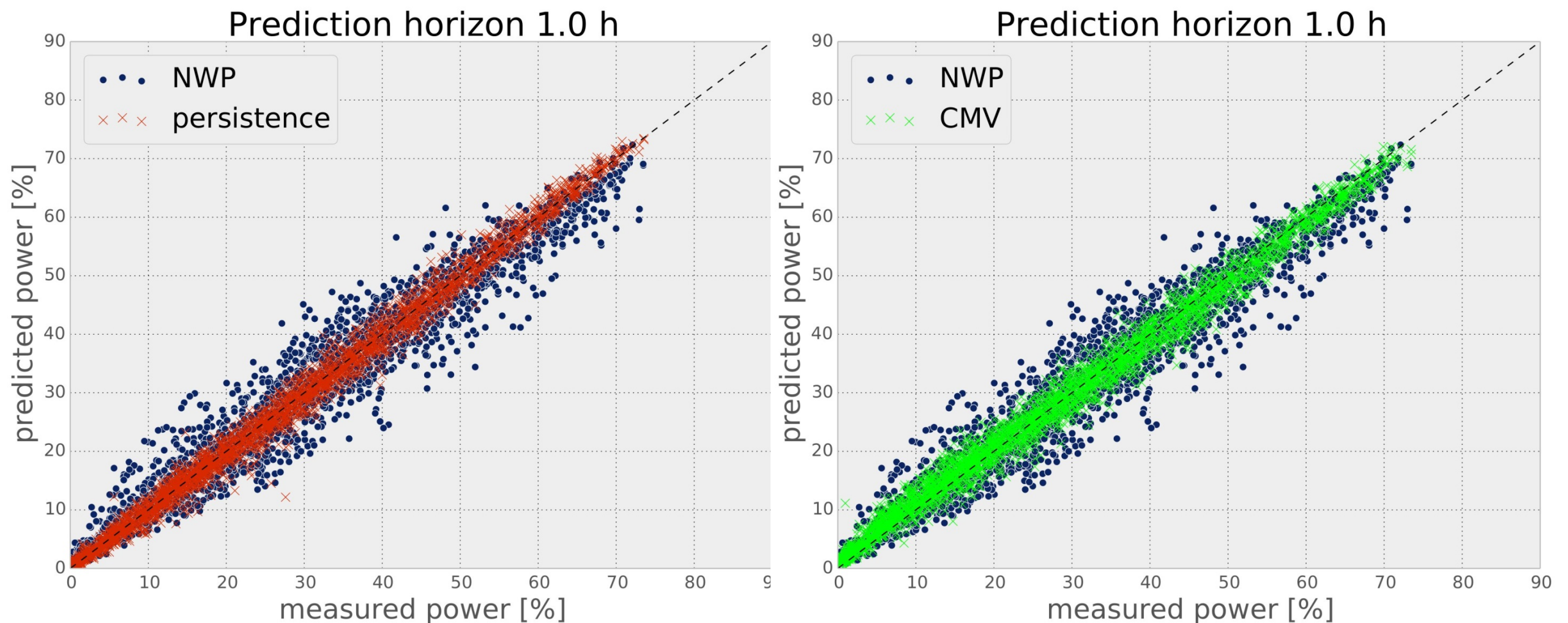
Different input data and models



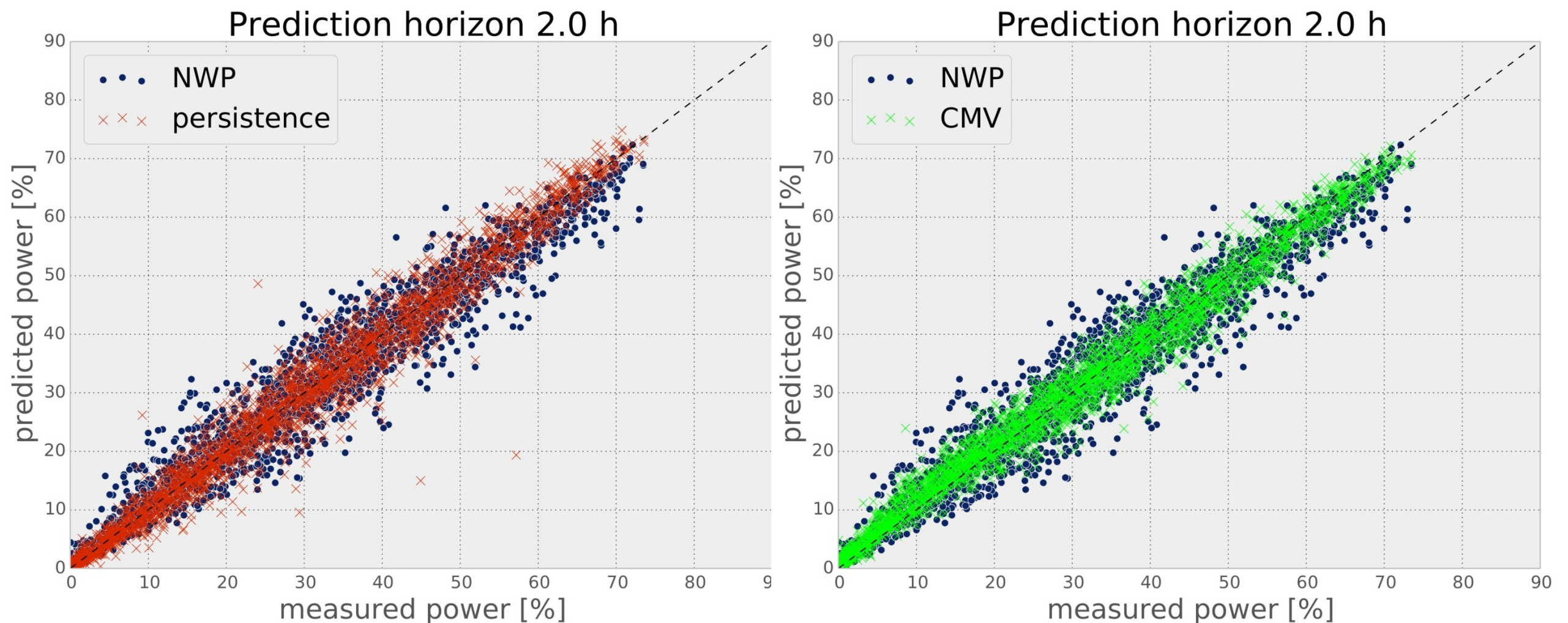
Regional forecasts: persistence, CMV and NWP based forecasts



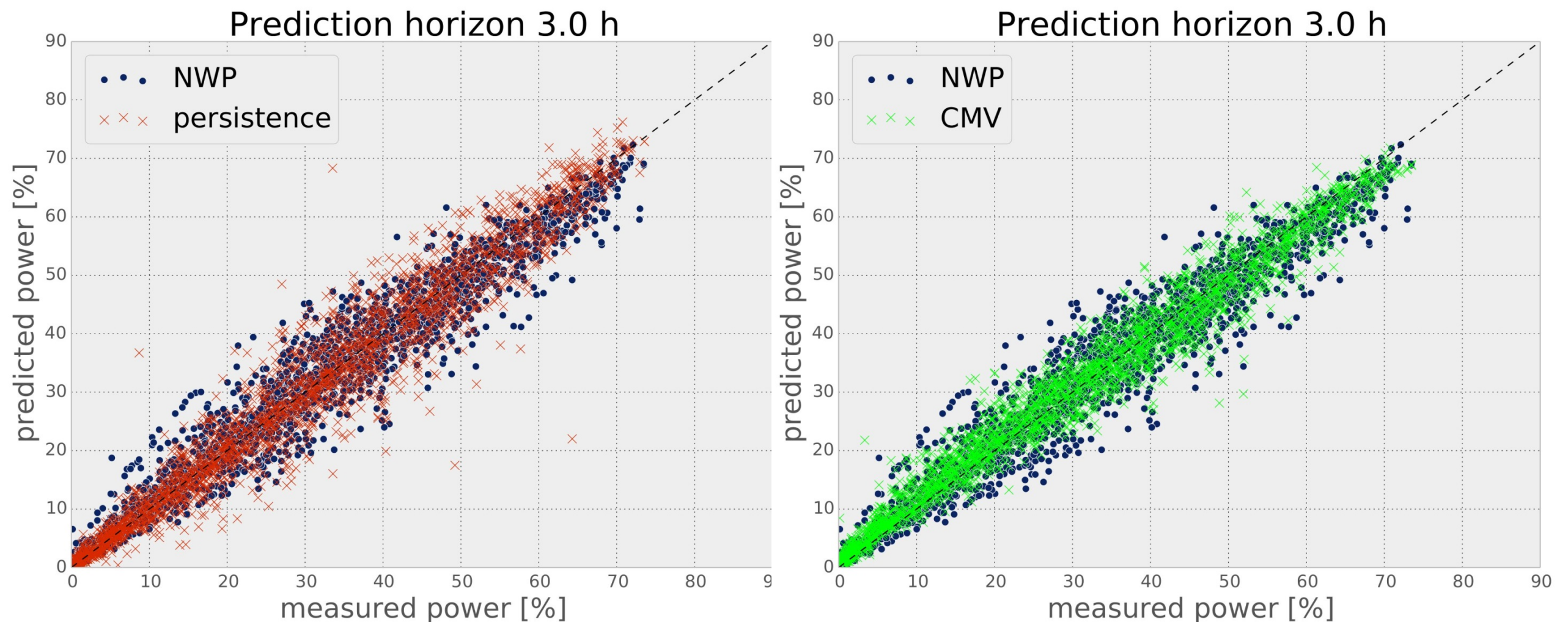
Regional forecasts: persistence, CMV and NWP based forecasts



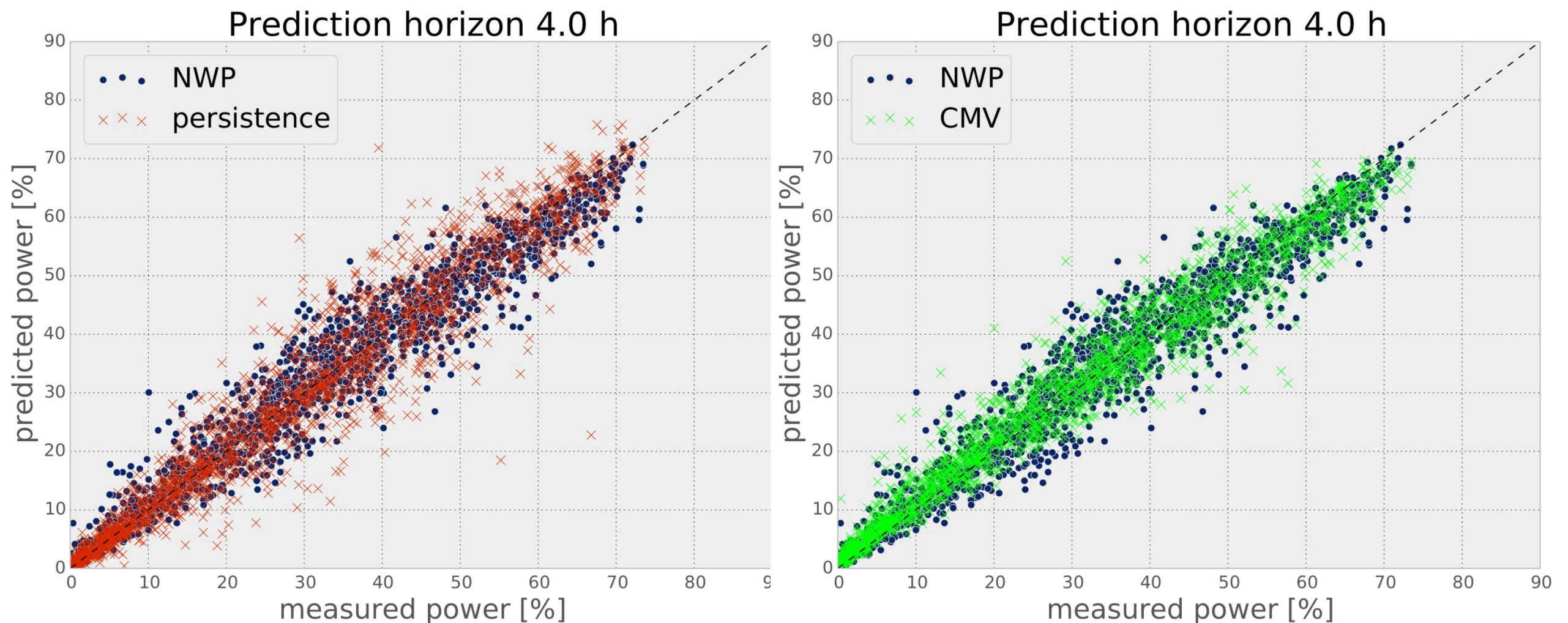
Regional forecasts: persistence, CMV and NWP based forecasts



Regional forecasts: persistence, CMV and NWP based forecasts

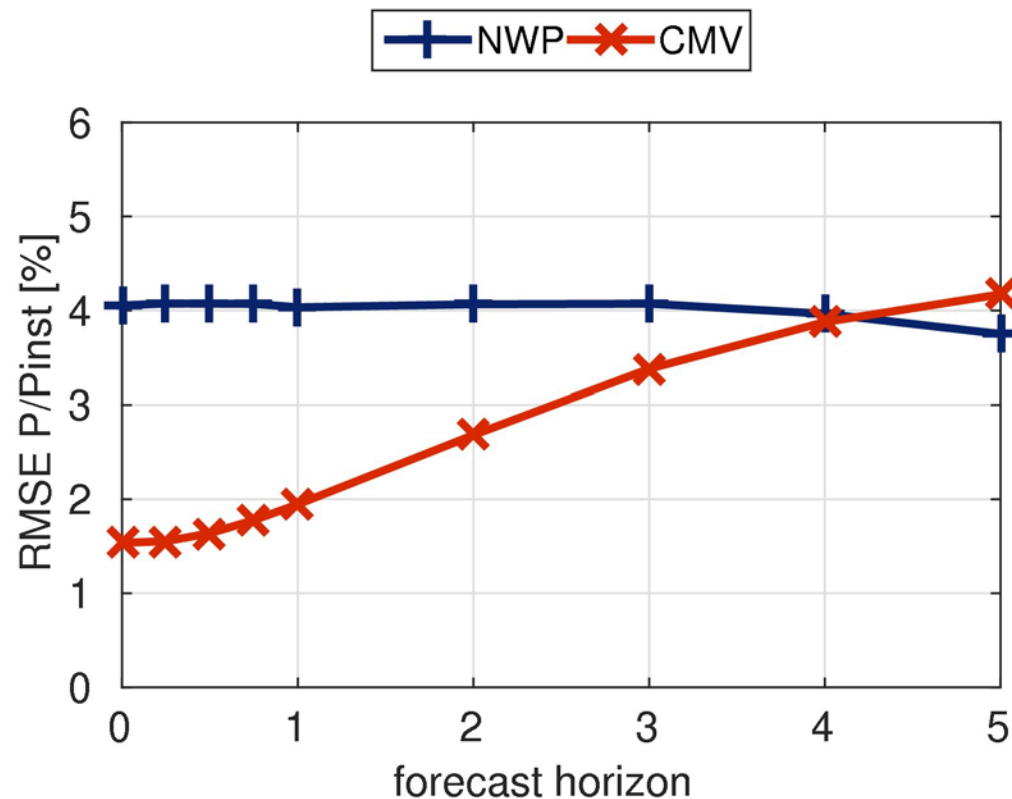


Regional forecasts: persistence, CMV and NWP based forecasts



Rmse in dependence of forecast horizon

German average



$$rmse = \sqrt{\frac{1}{N} \sum_{i=1}^N \left(\frac{P_{meas}}{P_{inst}} - \frac{P_{pred}}{P_{inst}} \right)^2}$$

15 minute values

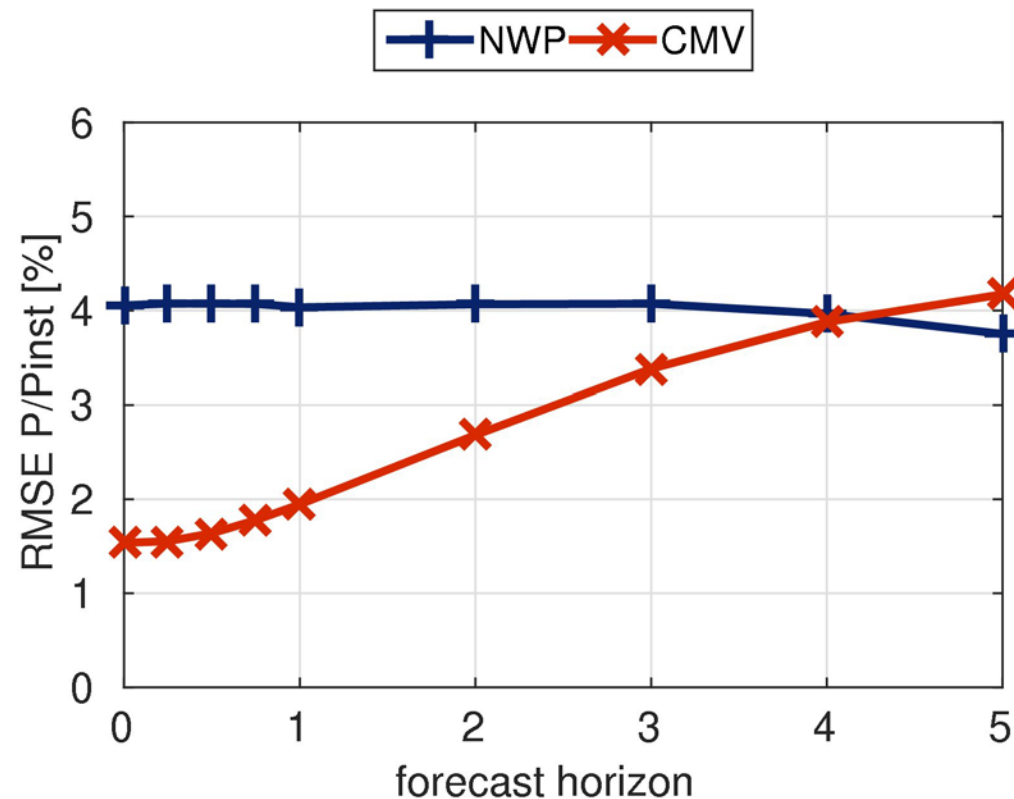
normalization to installed power P_{inst}

only daylight values, calculation time of CMV: $\text{sunel} > 10^\circ$

only hours with all models available included in dependence of forecast horizon

Rmse in dependence of forecast horizon

German average



$$rmse = \sqrt{\frac{1}{N} \sum_{i=1}^N \left(\frac{P_{meas}}{P_{inst}} - \frac{P_{pred}}{P_{inst}} \right)^2}$$

- ▶ CMV forecasts better than NWP based forecast up to 4 hours ahead

15 minute values

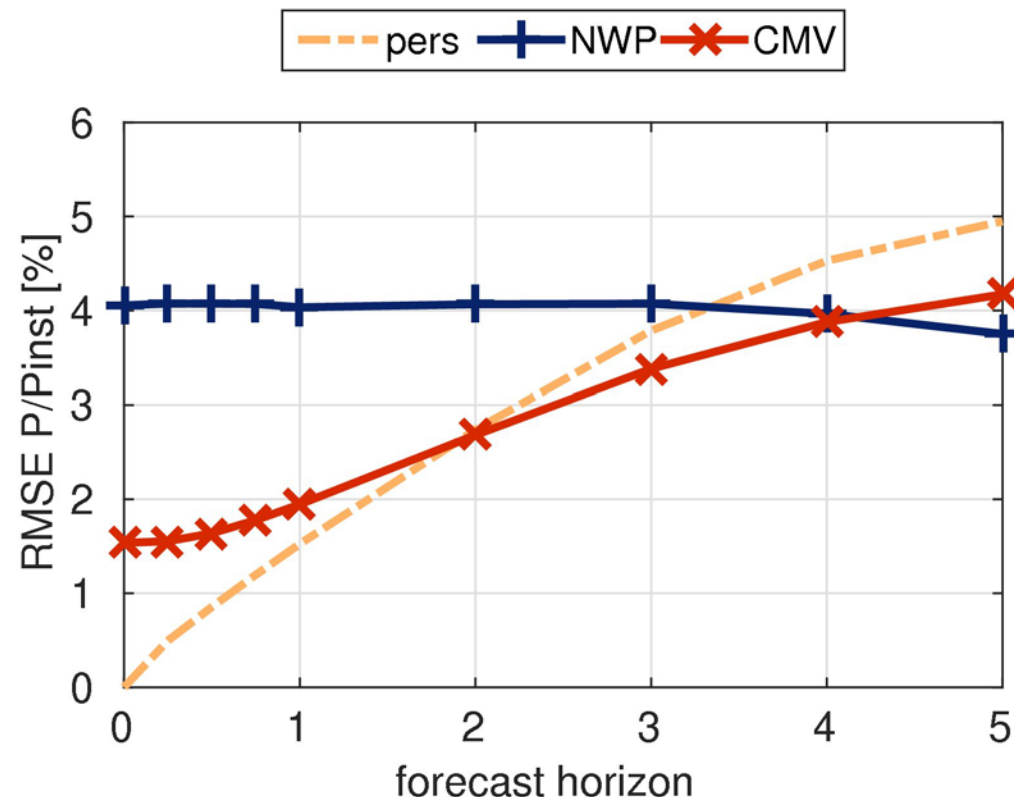
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Rmse in dependence of forecast horizon

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- ▶ CMV forecasts better than NWP based forecast up to 4 hours ahead
- ▶ persistence better than CMV forecasts up to 1.5 hour ahead

15 minute values

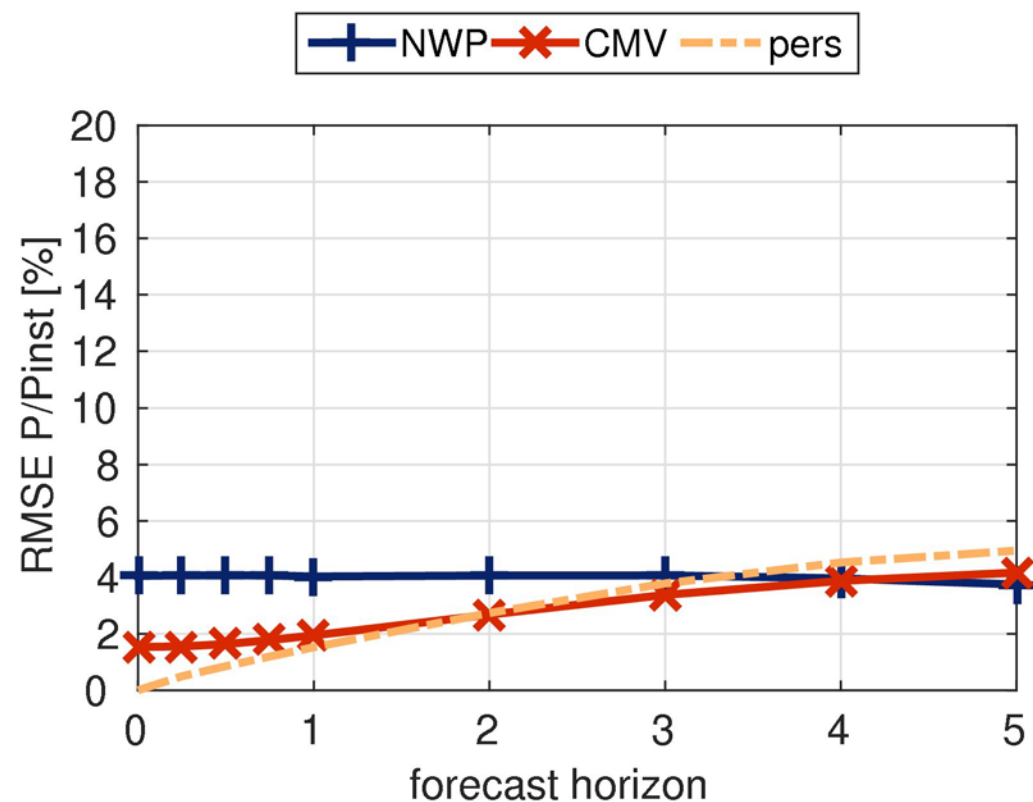
normalization to installed power P_{inst}

only daylight values, calculation time of CMV: $\text{sunel} > 10^\circ$

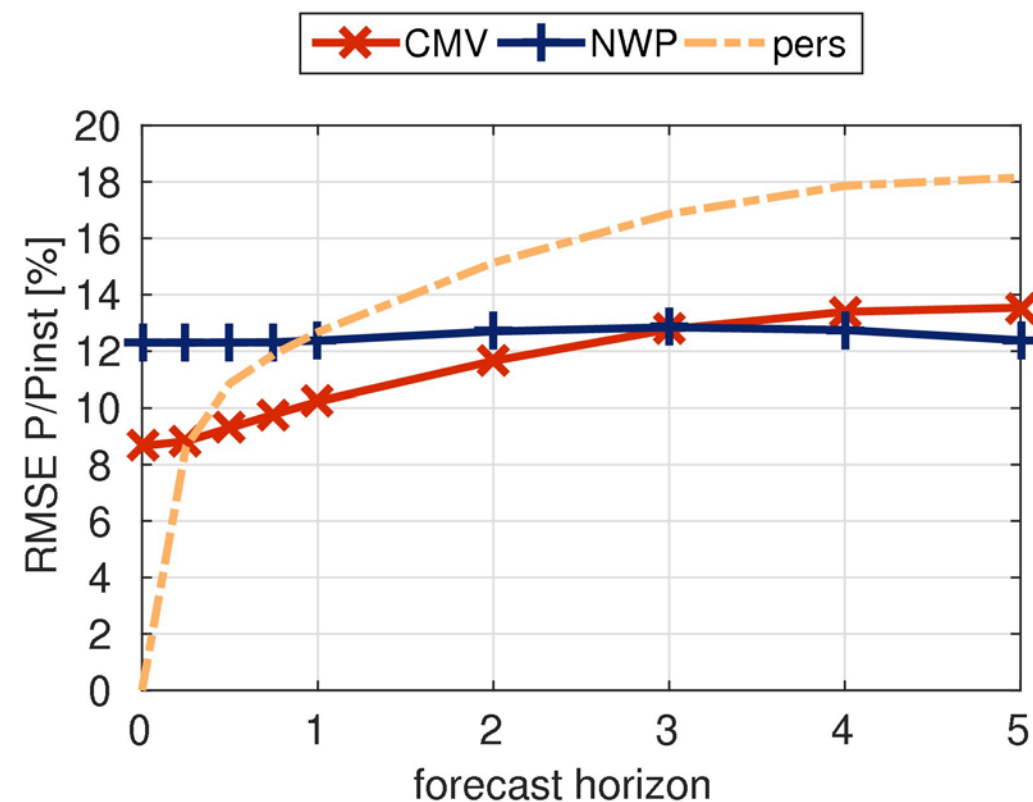
only hours with all models available included in dependence of forecast horizon

Rmse in dependence of forecast horizon

German average



single sites



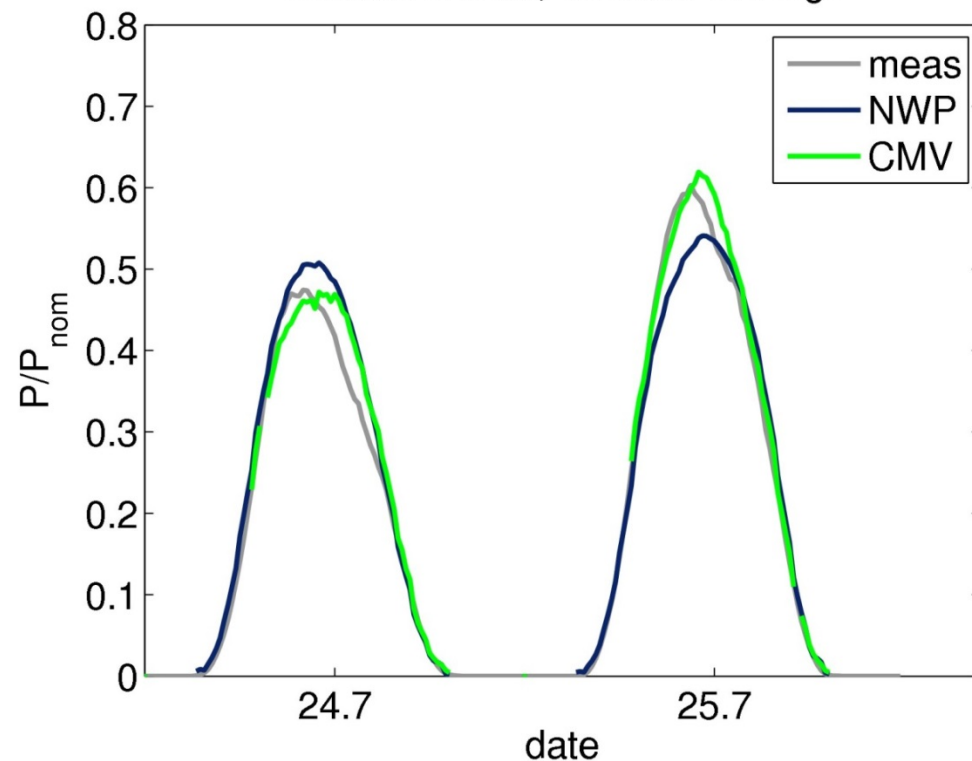
comparison of German average and single site forecasts:

- ▶ German average RMSE about 1/3 of single site RMSE for NWP forecasts

Rmse in dependence of forecast horizon

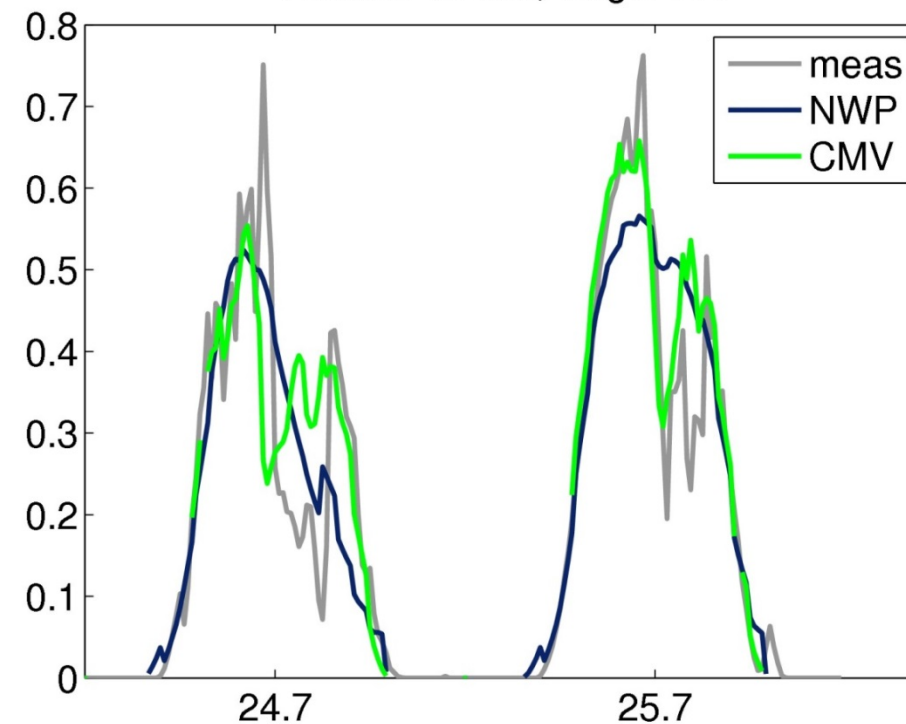
German average

variable clouds, German average



single sites

variable clouds, single site

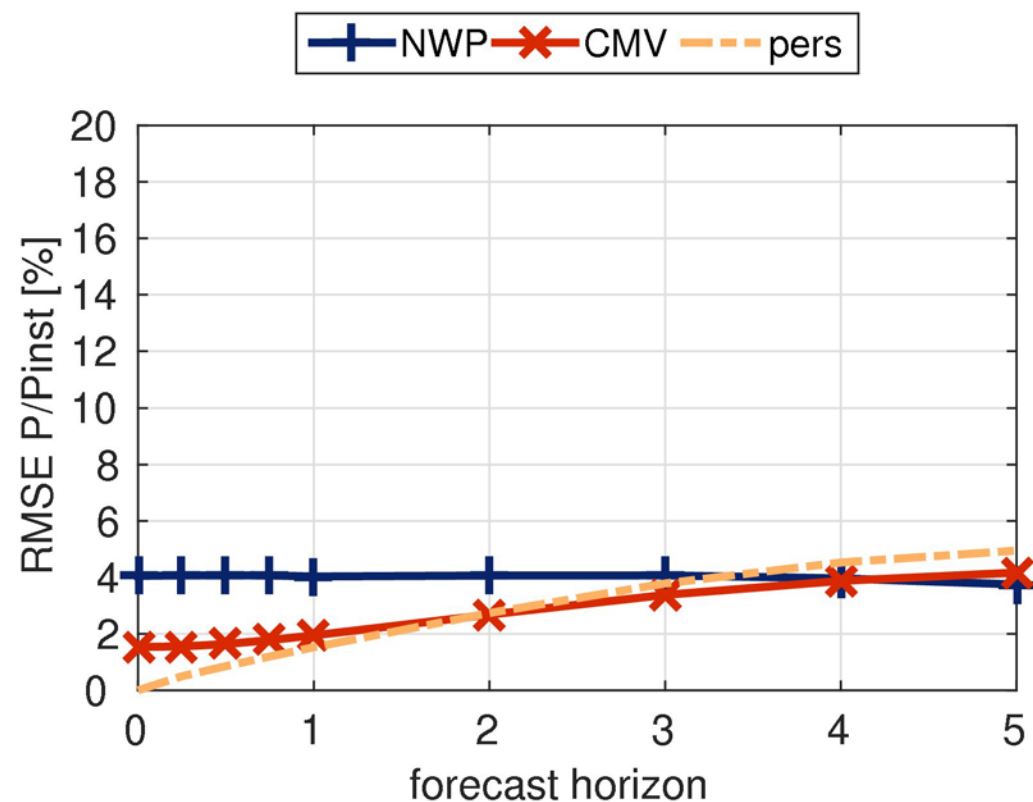


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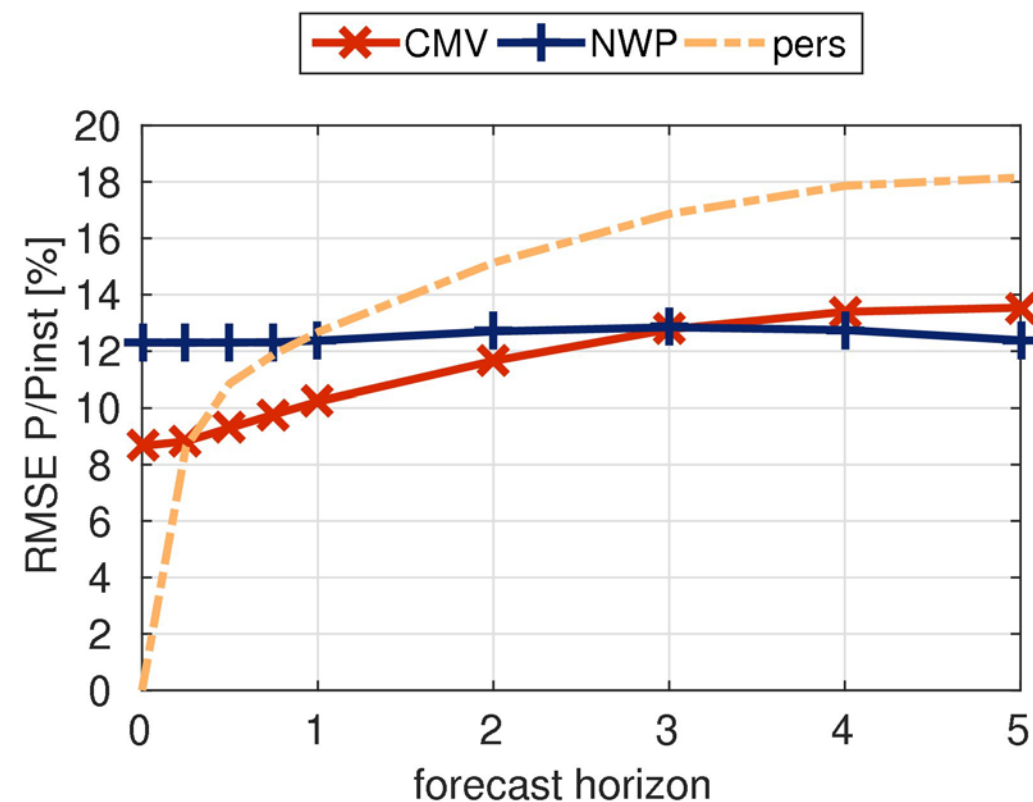
- ▶ German average RMSE about 1/3 of single site RMSE for NWP forecasts

Rmse in dependence of forecast horizon

German average



single sites

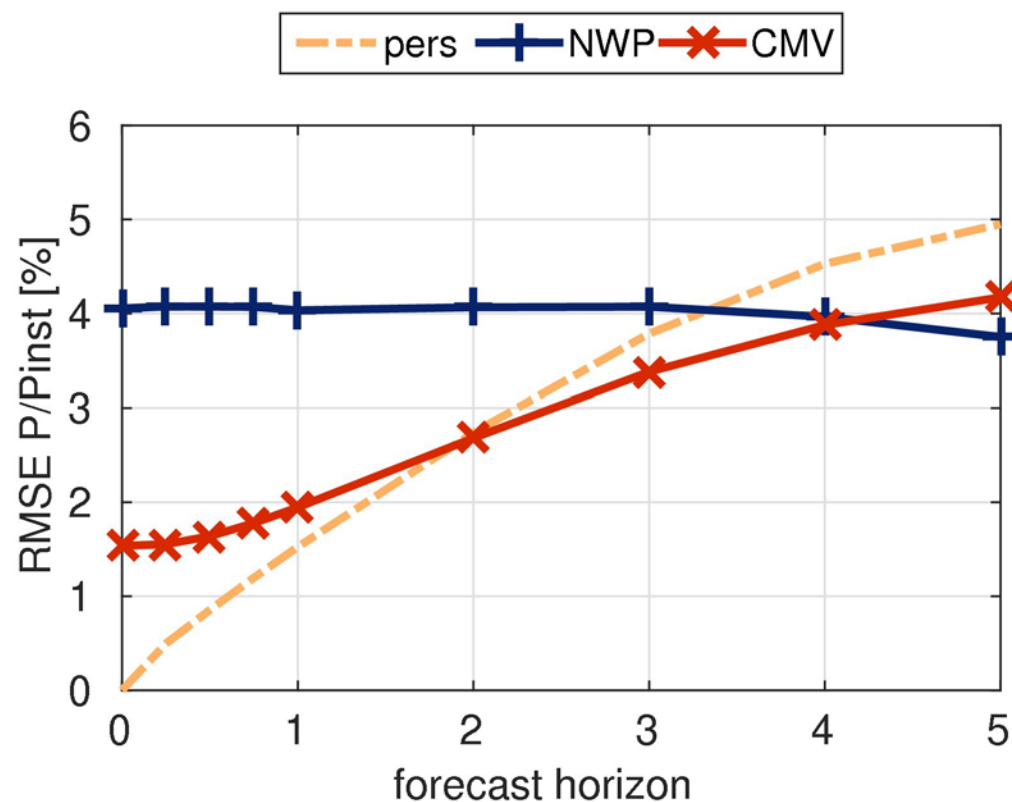


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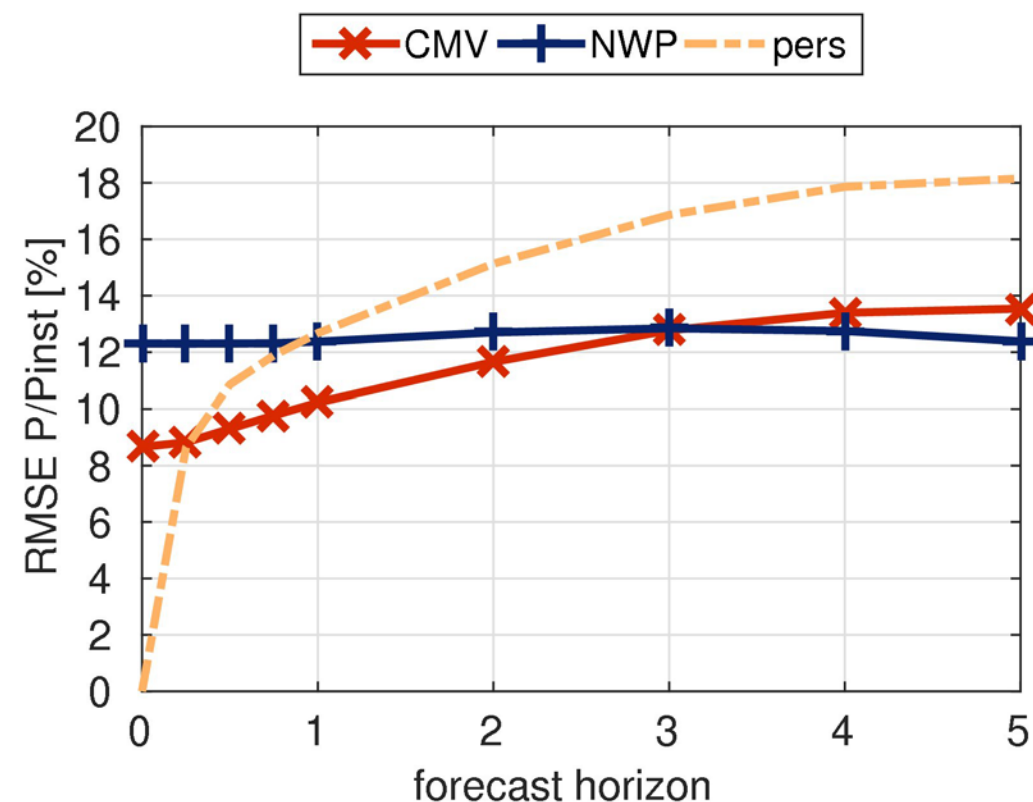
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Rmse in dependence of forecast horizon

German average



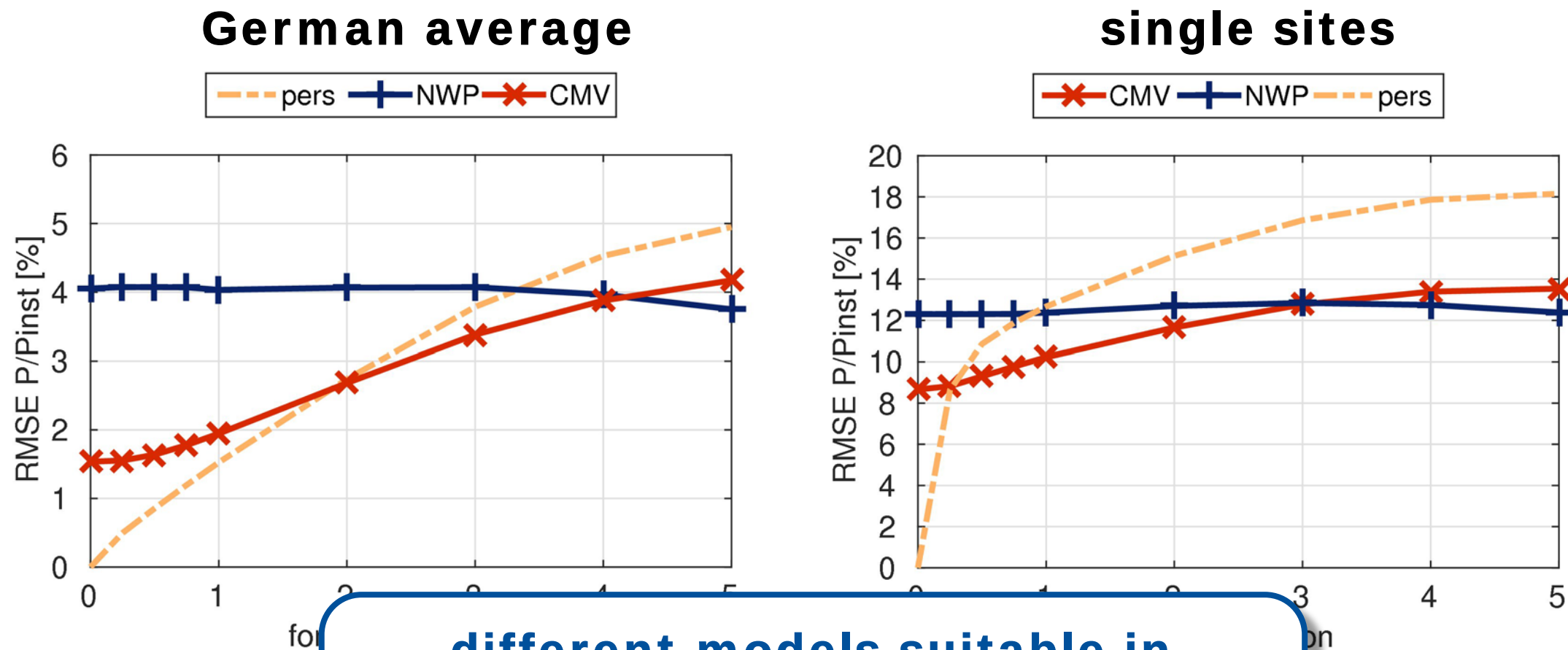
single sites



comparison of German average and single site forecasts:

- ▶ German average RMSE about 1/3 of single sites RMSE for NWP forecasts
- ▶ improvements with persistence and CMV larger for regional forecasts

Rmse in dependence of forecast horizon



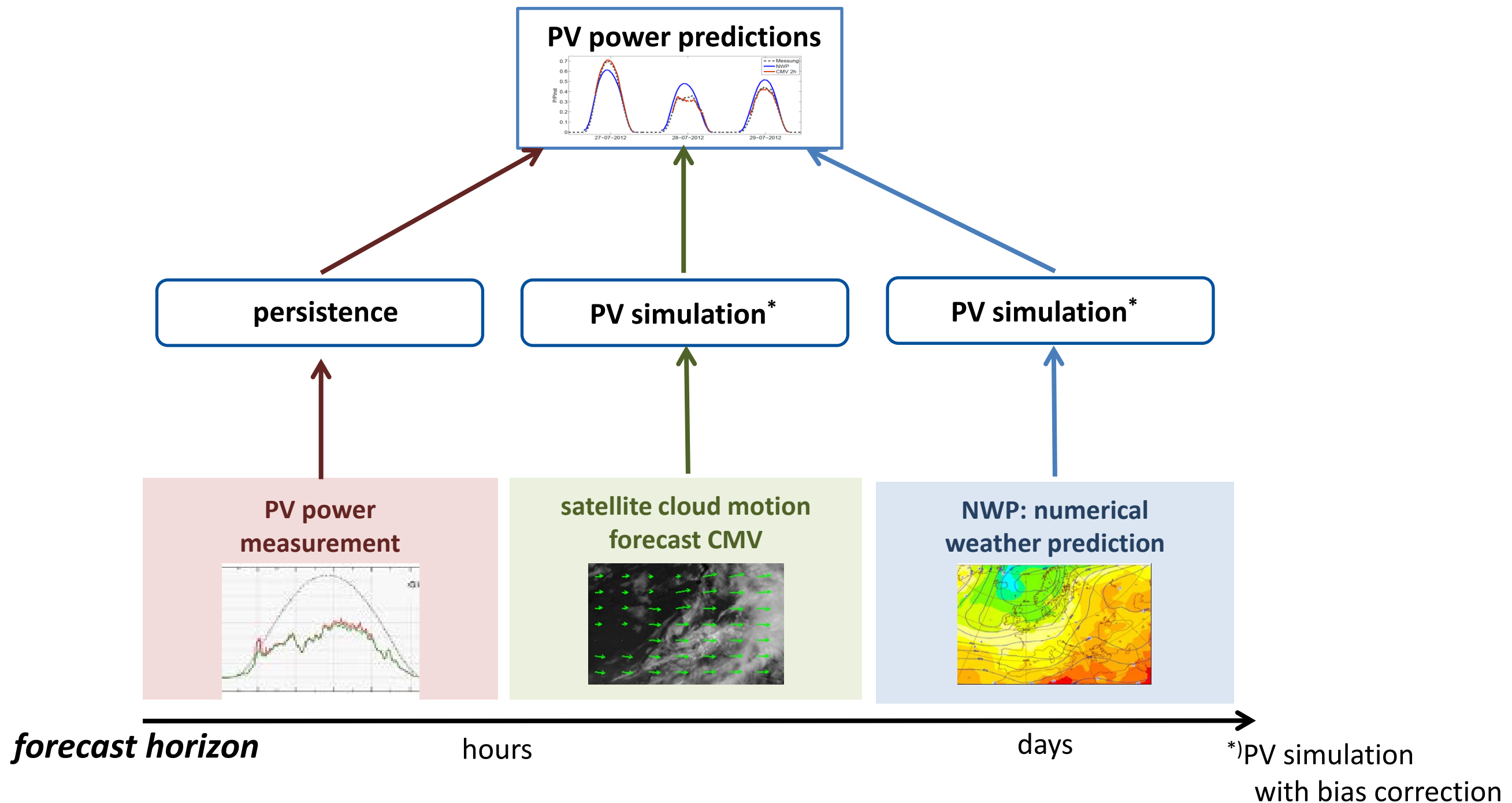
**different models suitable in
dependence of forecast horizon
and spatial scale**

comparis

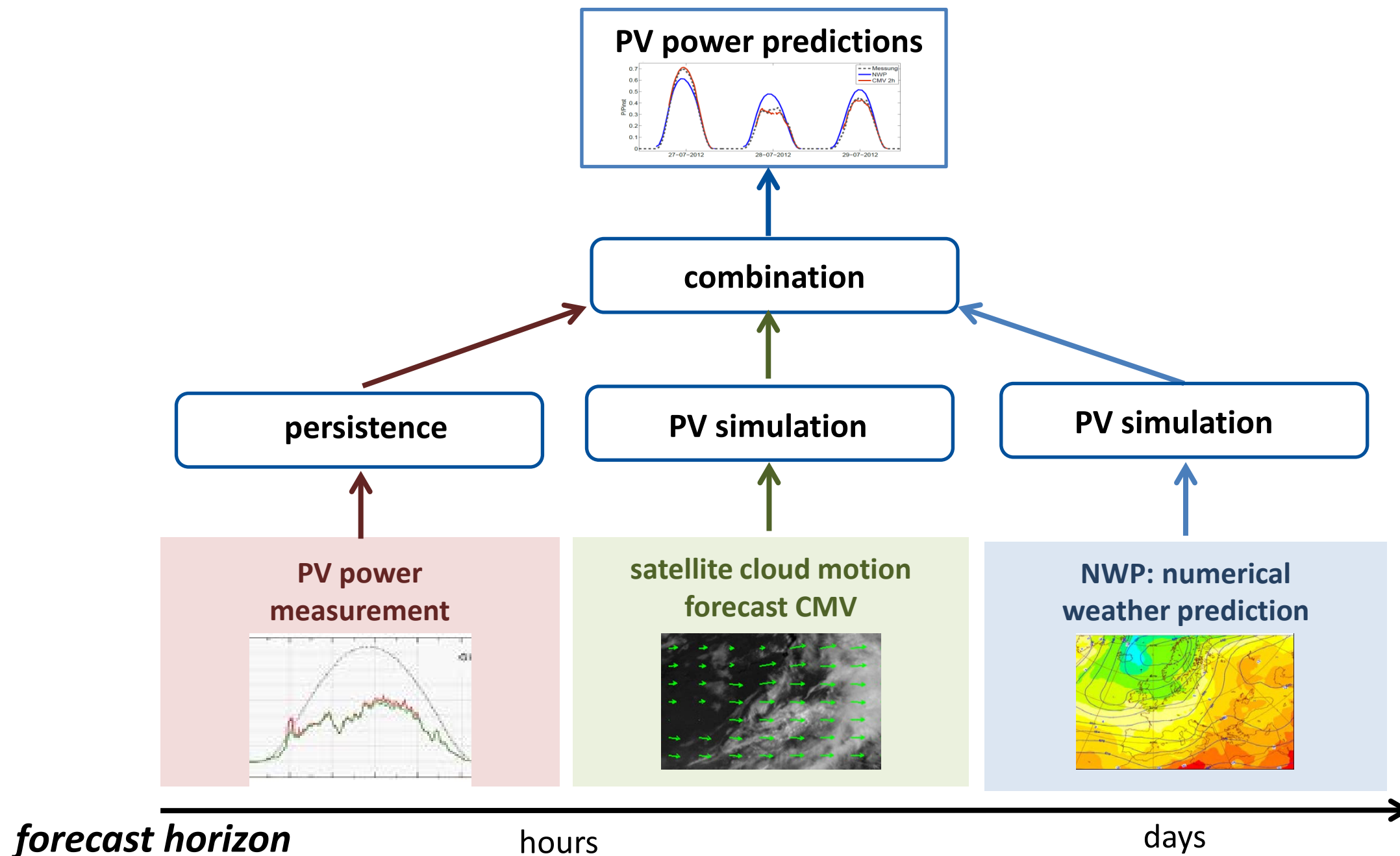
forecasts:

- ▶ German average RMSE about 1/3 of single sites RMSE for NWP forecasts
- ▶ improvements with persistence and CMV larger for regional forecasts

Different input data and models



Combination of different models



Combination of forecasting methods

combination of forecast models with linear regression:

$$P_{combi} = a_{NWP}P_{NWP} + a_{CMV}P_{CMV} + a_{persist}P_{persist} + a_0$$

coefficients a_{NWP} , a_{CMV} , $a_{persist}$, a_0 are fitted to measured data
in dependence of

- ▶ forecast horizon
- ▶ hour of the day

Combination of forecasting methods

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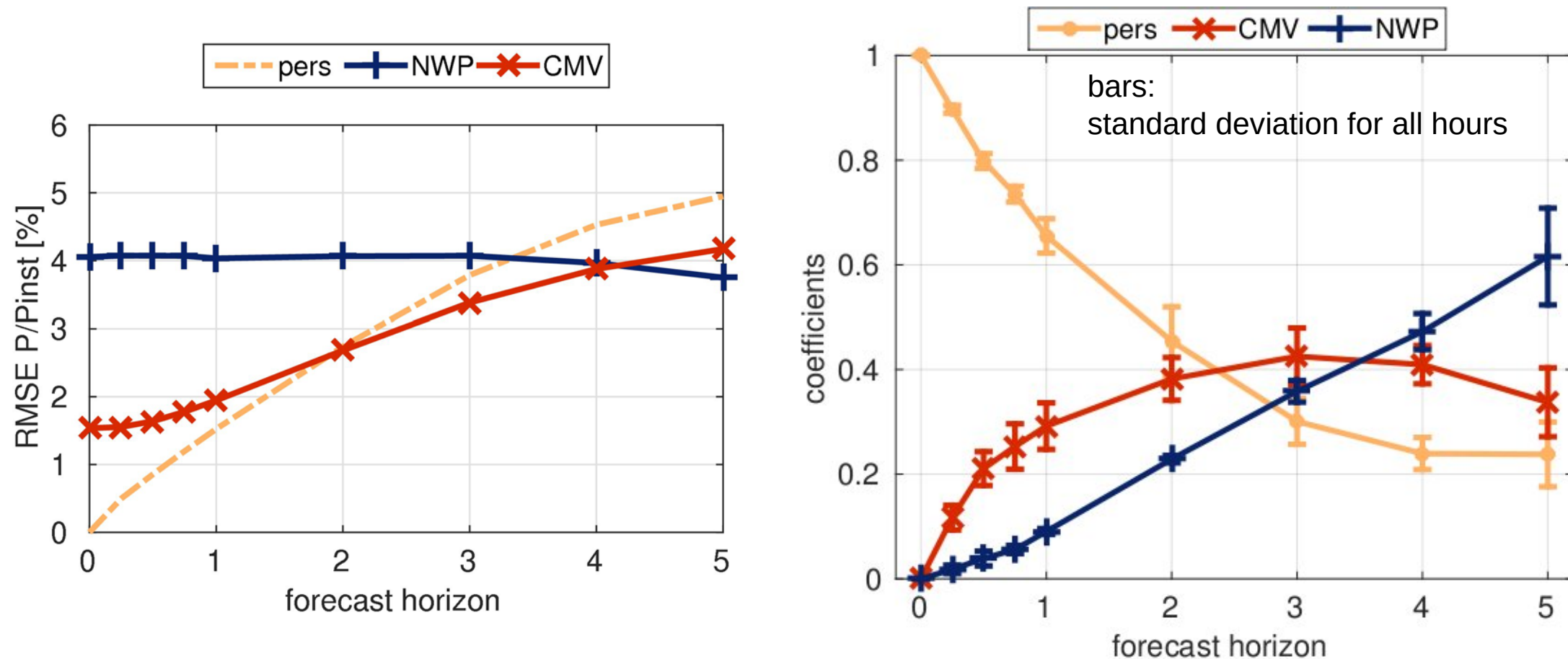
training data:

daily update with measurements of last 30 days

- ▶ for single site forecasts: each PV system separately
- ▶ for regional forecasts: average of sites

Regional forecasts

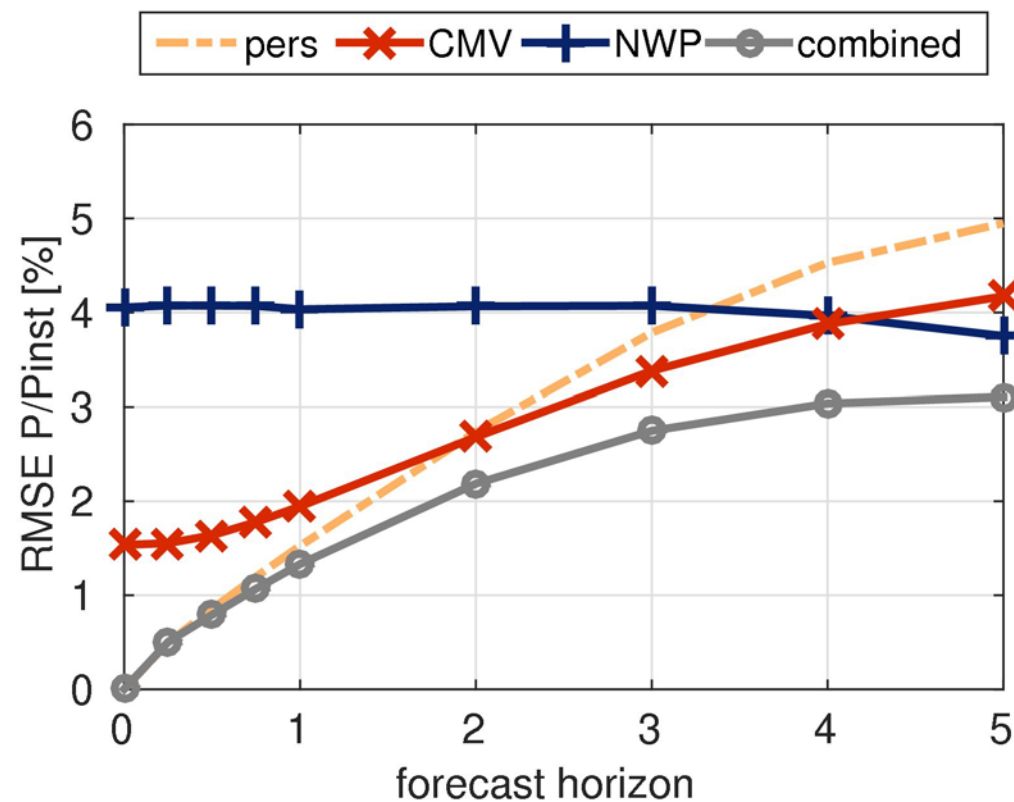
regression coefficients in dependence of **forecast horizon**



regression coefficients (weights) reflect horizon dependent forecast performance of different models

Regional forecasts

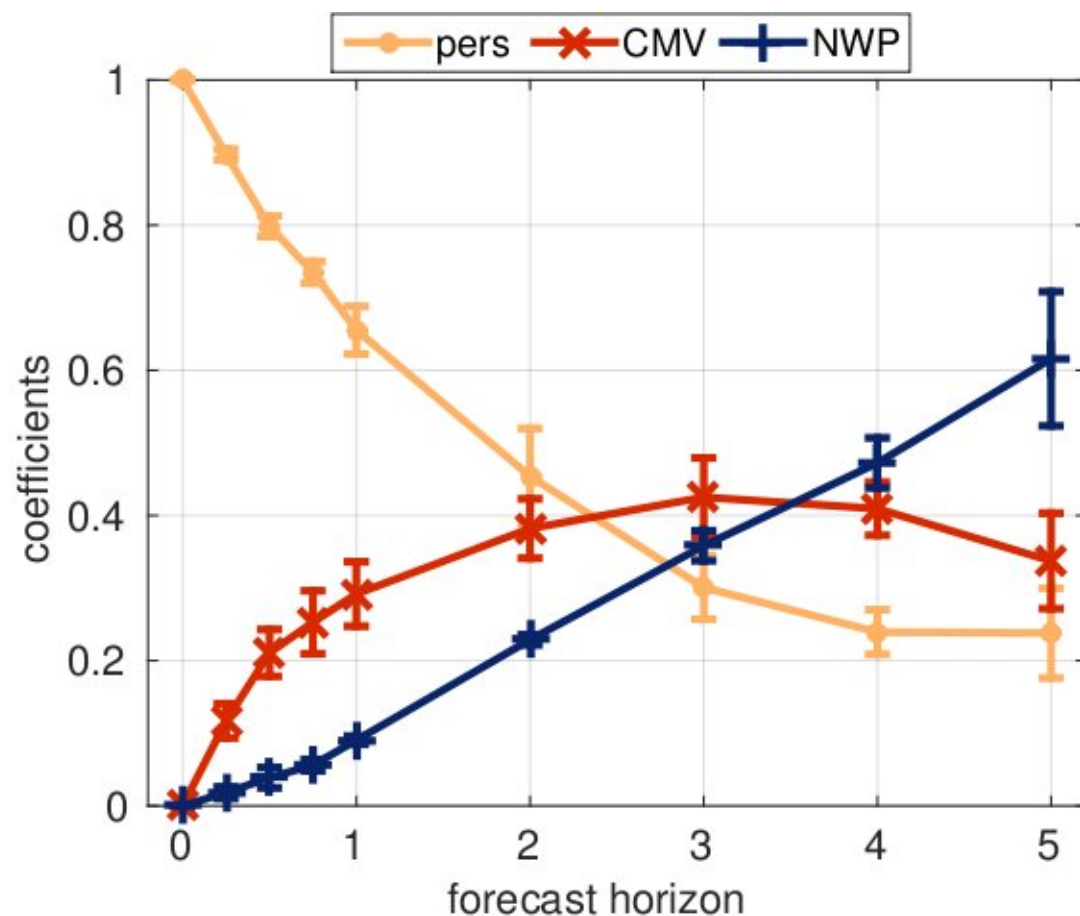
Rmse in dependence of forecast horizons



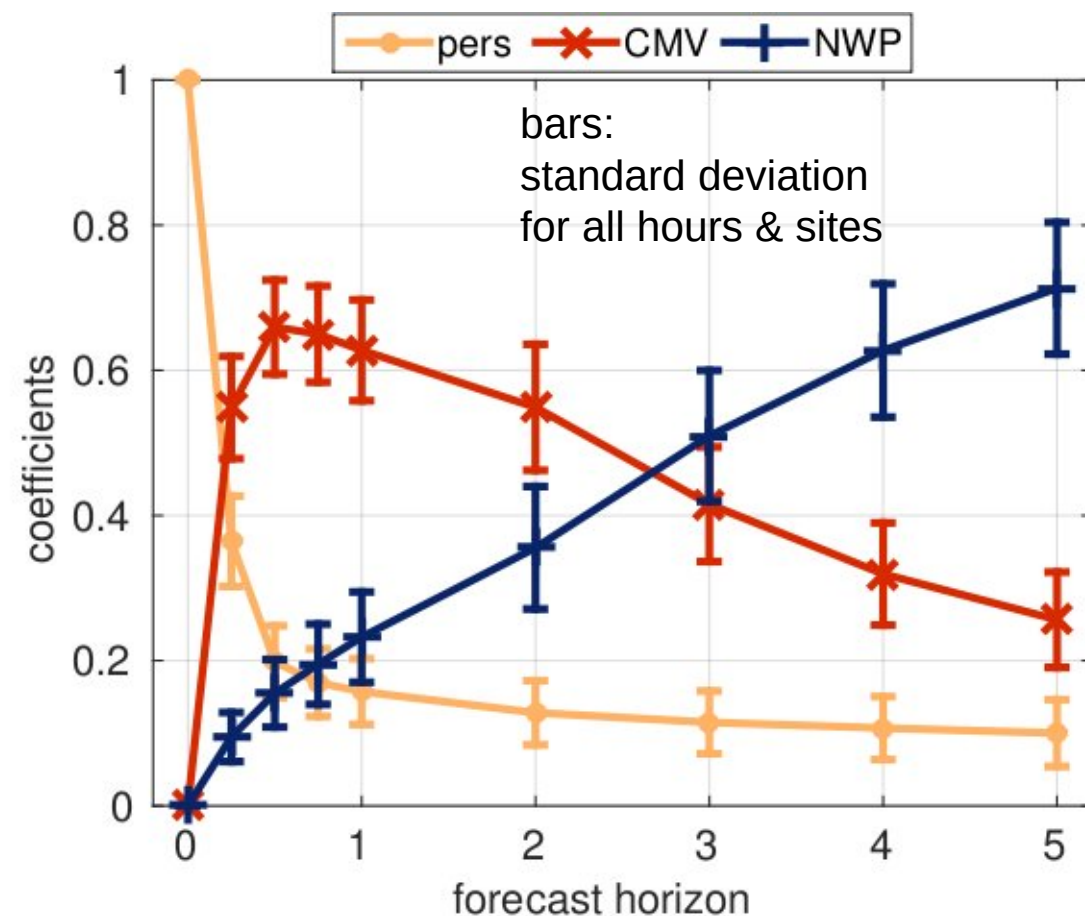
Considerable improvement with combined model over single model forecasts

Regression coefficients

German average



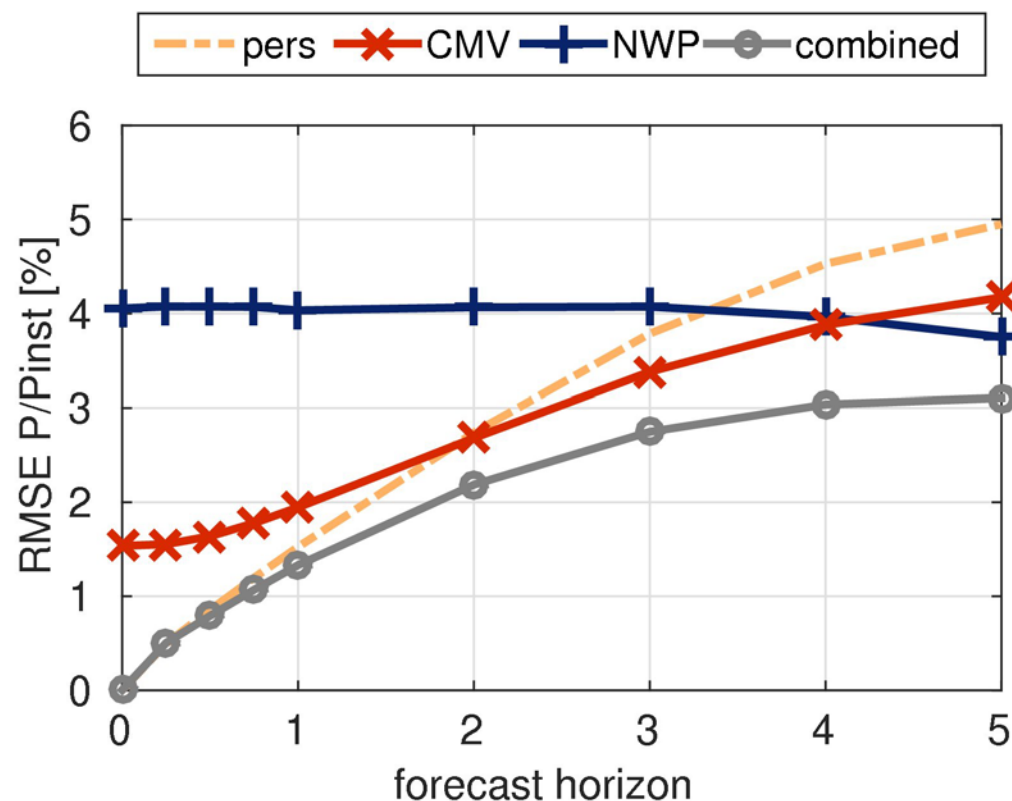
single sites



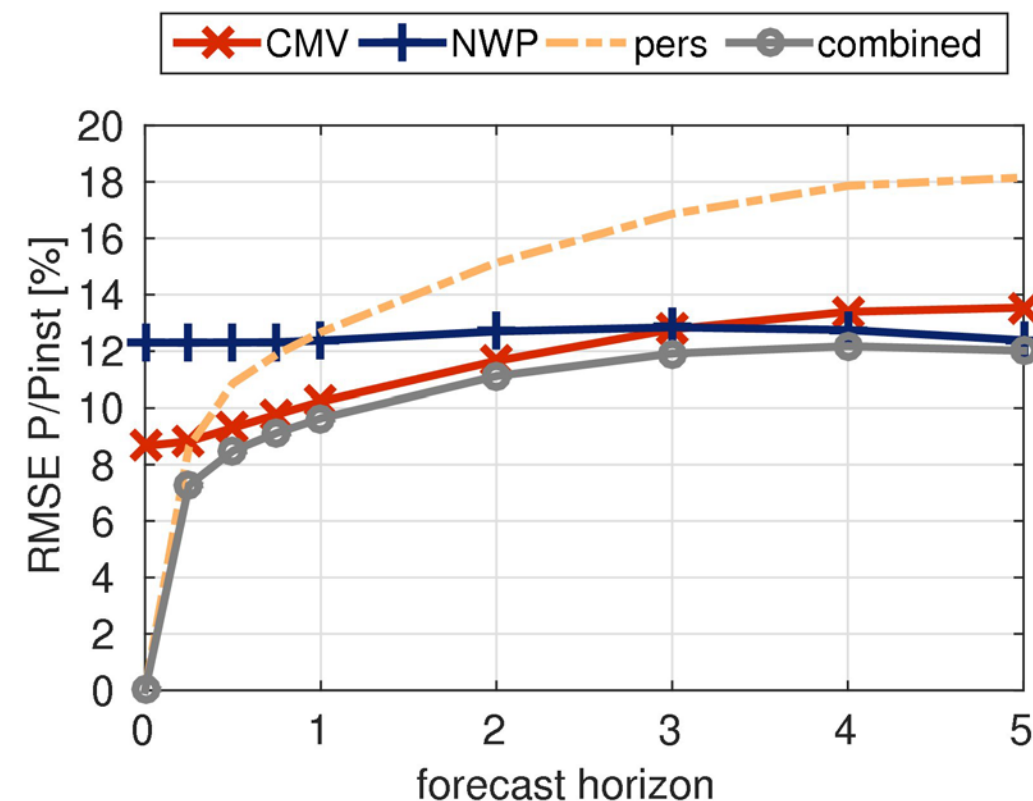
horizon dependent regression coefficients **different** for regional and single site forecasts

Rmse in dependence of forecast horizon

German average



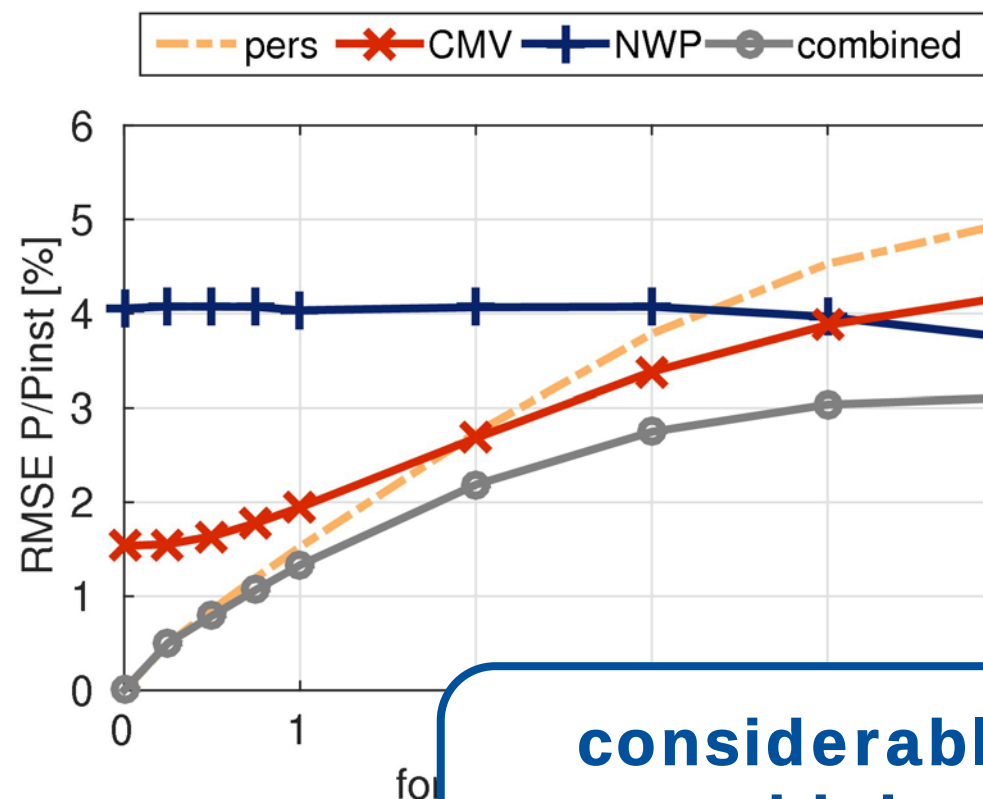
single sites



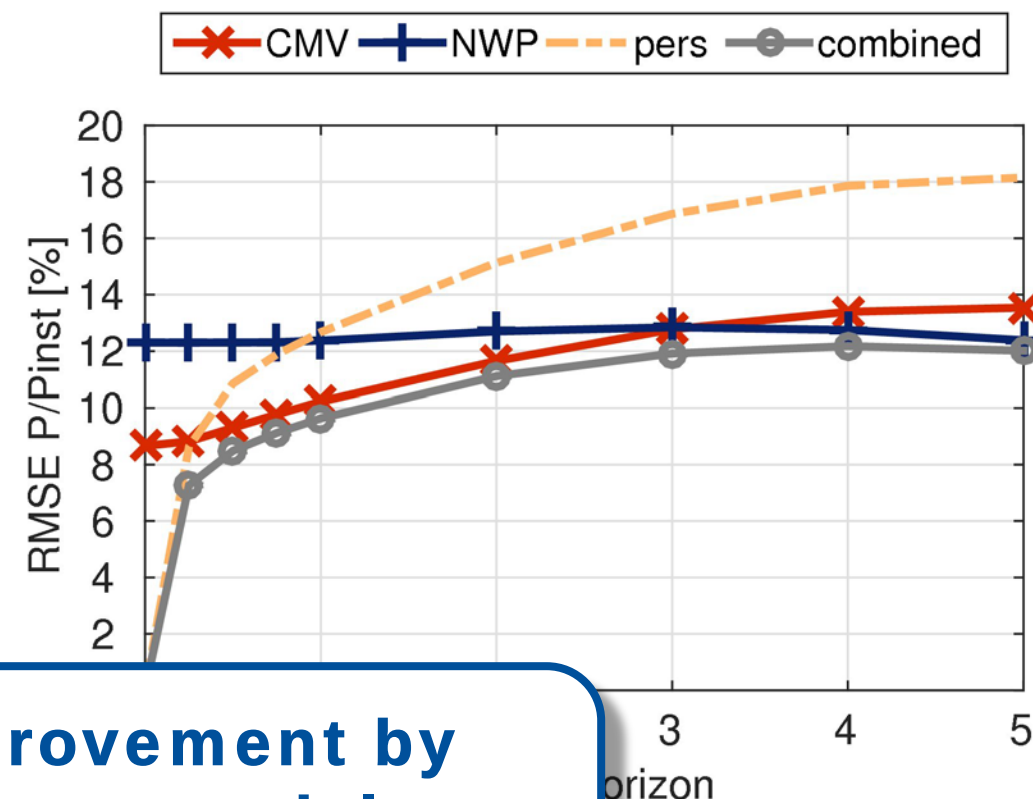
- ▶ forecast combination outperforms single model forecasts for all horizons
- ▶ improvements with combination larger for regional forecasts

Rmse in dependence of forecast horizon

German average



single sites



**considerable improvement by
combining different models
with statistical learning**

- ▶ forecast for all horizons
- ▶ improvements with combination larger for regional forecasts

Summary

- ▶ PV power prediction contributes to successful grid integration of more than 39 GW_{peak} PV power in Germany

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- ▶ PV power forecasts based on satellite data (CMV) significantly better than NWP based forecasts up to 4 hours ahead

Summary

- ▶ PV power prediction contributes to successful grid integration of more than 39 GW_{peak} PV power in Germany
- ▶ PV power forecasts based on satellite data (CMV) significantly better than NWP based forecasts up to 4 hours ahead
- ▶ significant improvement by combining different forecast models with PV power measurements, in particular for regional forecasts

Thank you for your attention!

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für Wirtschaft
und Energie

aufgrund eines Beschlusses
des Deutschen Bundestages

Gefördert durch das
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