Conference: "The role of market design for the achievement of the 20% Renewables Target", Brussels, 10.06.2010.

The uncertainty of wind output – how it declines during the day.

State of the art and research priorities in wind power forecasting

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1. Introduction

- 2. General principles of wind power forecasting
- 3. Typical accuracy of state of the art models
- 4. R&D projects & priorities



Introduction



- Reliable large-scale integration
 - Economic and secure management of a power system
 - Competitiveness of wind energy in a liberalised electricity market



Introduction

- Wind power forecasting for the next minutes to days is recognised today as a necessary tool for 'smooth' wind integration
- However, forecasting wind power is a complex problem.
 Some of the reasons...



Wind is highly variable by nature... (example: wind production of a wind farm during a month)



The wind turbine characteristic curve introduces important non-linearities.



The principle



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Types of forecast errors



Factors affecting accuracy

- Prediction horizon
- ✓ Time of the year (climatic conditions)
- ✓ Terrain complexity
- ✓ Types of inputs/types of models
- ✓ Spatial smoothing effect
- Level of predicted power and speed

Factors affecting accuracy: Prediction horizon

Average error increases with horizon

Case Study: wind farm in Ireland



Factors affecting accuracy: Time of the year

• Error varies throughout the year

Persistence **F-NN model NMAE [%]** Nov. Dec. Jan. Feb. Mar. Apr. May Jun. Jul. Aug. Total Oct. Month of the year and Total

Case Study: wind farm in Ireland



Factors affecting accuracy: Spatial smoothing effect

Example :

Prediction of 2200 MW in Jutland area in DK

- SCADA measurements only from 23 WFs (~200 MW).
- Differenced measurements of the total production.
- Prediction of 23 WFs output
- Then upscale to the total area.





Typical performance

MEASURE :

 Normalised Mean Absolute Error (NMAE)

TYPICAL VALUES :

 4% -18% for a single wind farm in complex terrain. Comparison of 3 statistical (M1-M3) and 2 physical (M4-M5) models

Case Study: wind farm in Ireland



Much lower error for regional/national forecasting due to spatial smoothing effect for horizons >4-6 hours. ~5-10% for 24-48 hours.



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Context – Research in Wind Power Forecasting

• Considerable research carried out in the last 20 years HIGHLIGHT PROJECTS :

ANEMOS : FP5, 2002-2006

ANEMOS.plus : FP6, 2008-2011





R&D priorities*

- Better estimation of uncertainty Probabilistic forecasting Ensembles
- Prediction of "extreme events" (i.e. ramps)
- Research in meteorology oriented to wind power forecasting
- Methods based on geographically distributed measurements
- Methods for alarming and warning
- Prediction risk indices
- Scenarios
- Reliable prediction systems





R&D priorities*

 Decision making based on forecasts and uncertainty estimation Advanced tools are demonstrated in Anemos.plus for :

Probability density

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isTech

look ahead time [h]

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- reserves estimation,
- power system scheduling,
- coordination of wind with storage,
- congestion management,
- strategic bidding in markets

Thank you for your attention



