

Danmarks Tekniske Universitet





03 December 2013 «WIRE» COST Action, benchmark results

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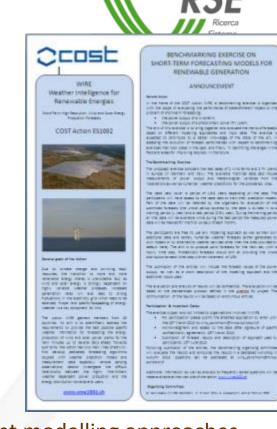


Objectives & Motivation

In the frame of WG1 of WIRE a benchmarking exercise was organised:

The **main objective** is to evaluate the performance of state-of-the-art models on the problems of short-term forecasting of:

- the power output of a wind farm,
- the power output of a photovoltaic power (PV) plant.

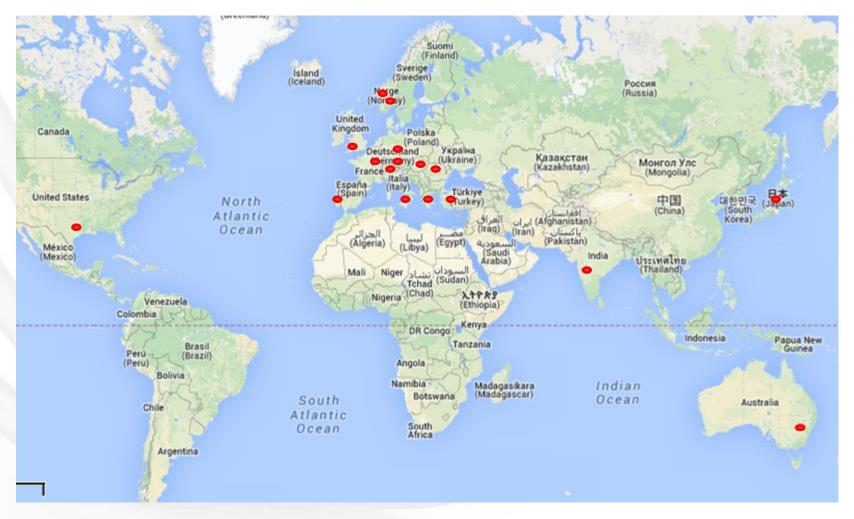


The exercise evaluates the merits of forecasts based on different modelling approaches and input data (data from 2 wind farms and 2 PV plants considered).

It contributes to a better knowledge of the state of the art, and its evolution through time (comparison with results from past exercises).

It permits to identify challenges in the field and areas for improving accuracy in the future.





18 Total participants

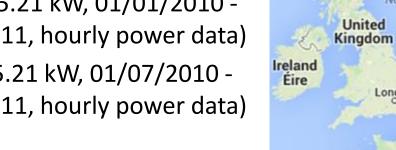
General Data Description

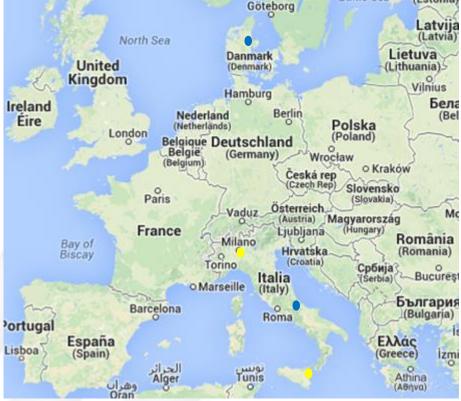
2 solar PV farms (Italy)

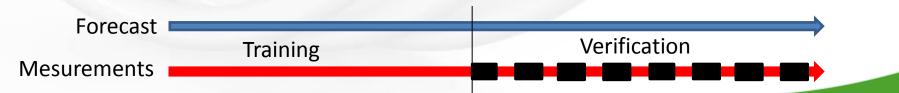
- **Catania** (5.21 kW, 01/01/2010 -31/12/2011, hourly power data)
- Milano (5.21 kW, 01/07/2010 -31/12/2011, hourly power data)

2 wind farms (Italy, Denmark)

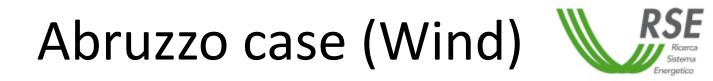
- Abruzzo (104 MW, complex terrain, 01/01/2010-31/12/2011 hourly power data)
- Klim (21 MW, flat terrain, 01/01/2001-31/12/2002 hourly power data)

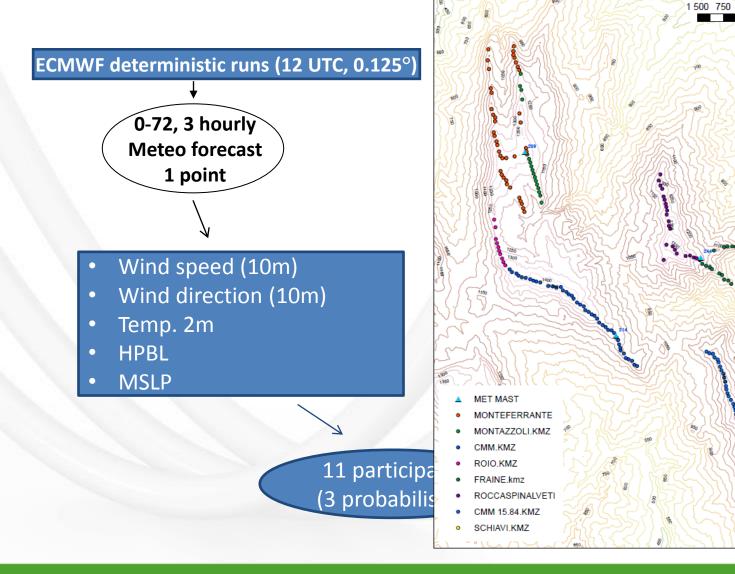








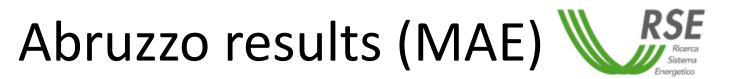


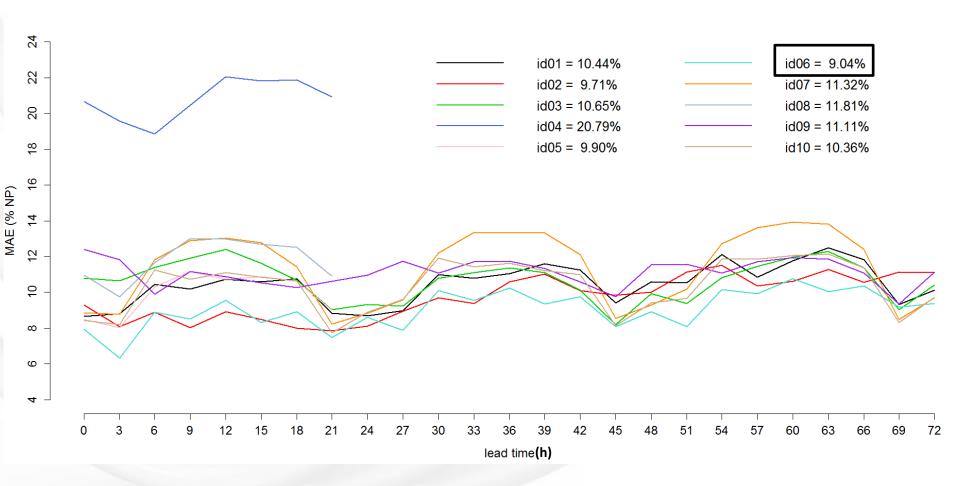


1 500 Meters

8 388

0





Abruzzo results (deterministic)



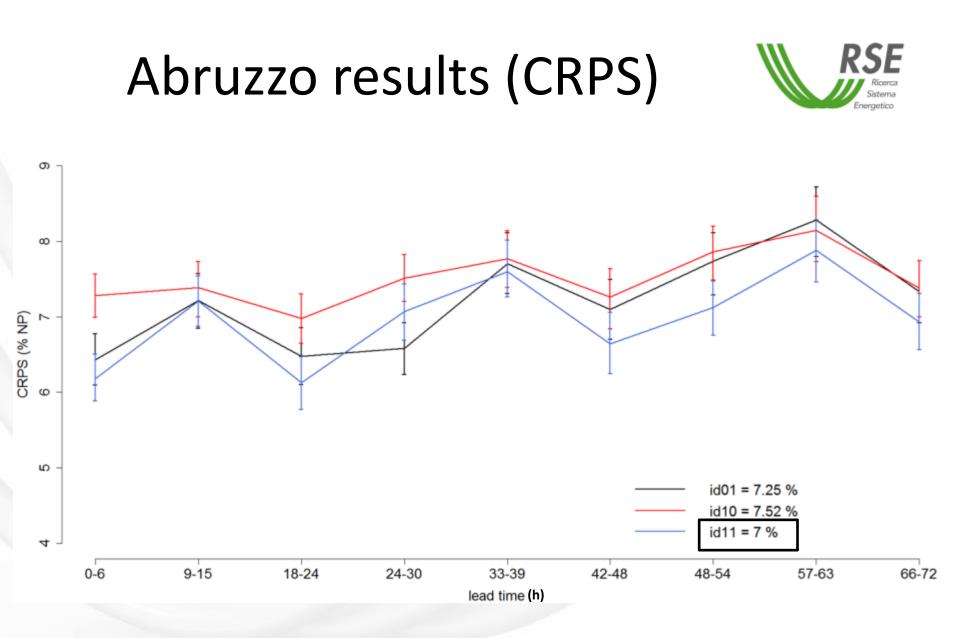
Ranking

- 1. <u>id06</u> (9.04 % MAE)
- 2. id02 (9.71 % MAE)

Diebold-Mariano statistic

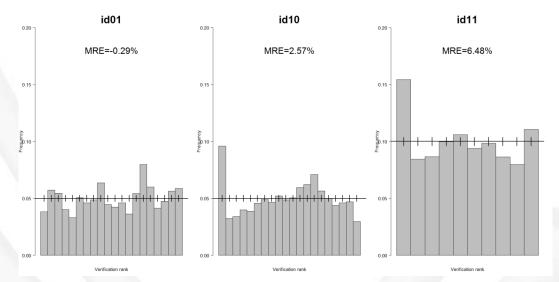
- H0: the two models have the same forecast accuracy
- H1: 2nd model is less accurate than the 1st one
- p-value=<u>3.91*10⁻⁶</u>

- Meteorological model: ECMWF provided by COST
- Post-processing with an Artificial Neural Network



Abruzzo results (probabilistic)





Ranking (probabilistic)

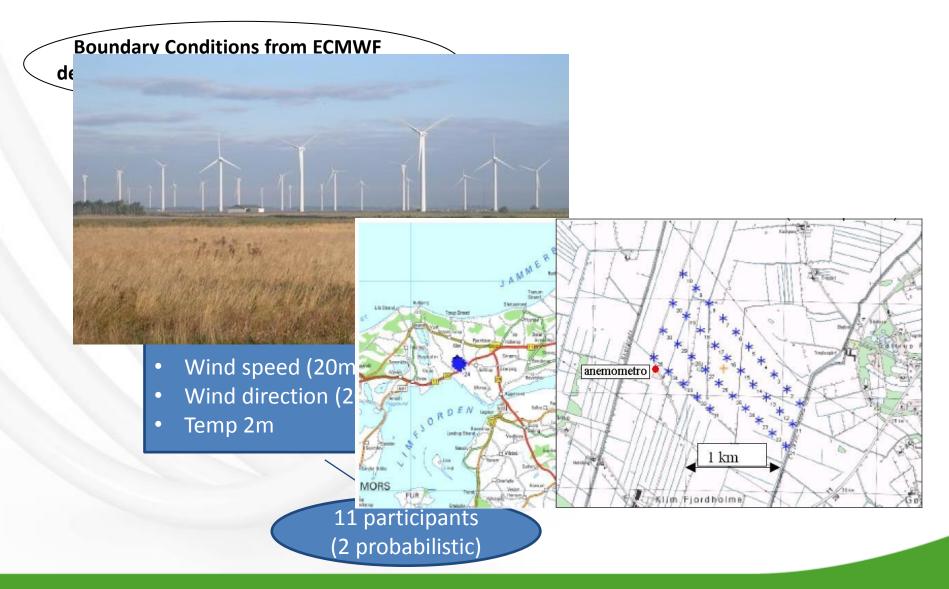
- **1.** <u>id11</u> (7.0 % CRPS)
- 2. id01 (7.25 % CRPS)

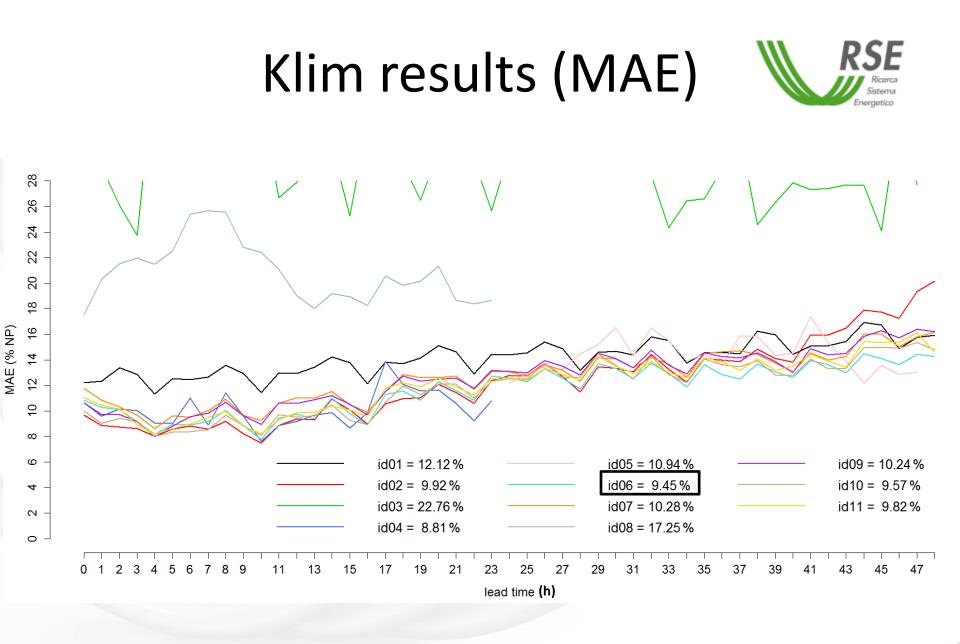
Method used by id11

Local quantile regression with wind speed and wind direction as predictors.

KLIM case (Wind)







Klim results (deterministic)



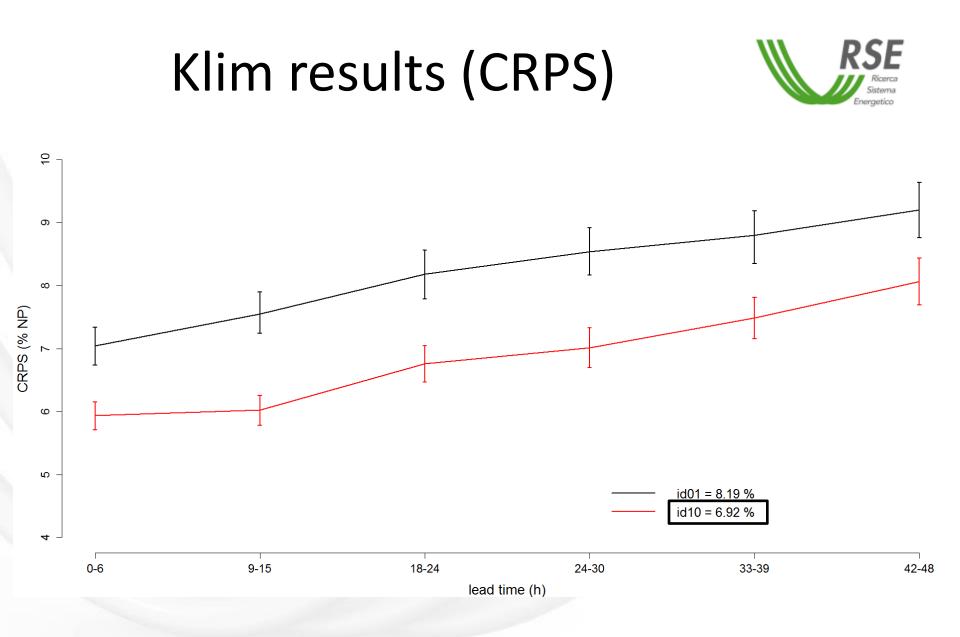
Ranking

- 1. <u>id06</u> (9.45 % MAE)
- 2. id10 (9.57 % MAE)

Diebold-Mariano statistic

- H0: the two models have the same forecast accuracy
- H1: 2nd model is less accurate than the 1st one
- p-value=<u>4.68*10⁻⁷³</u>

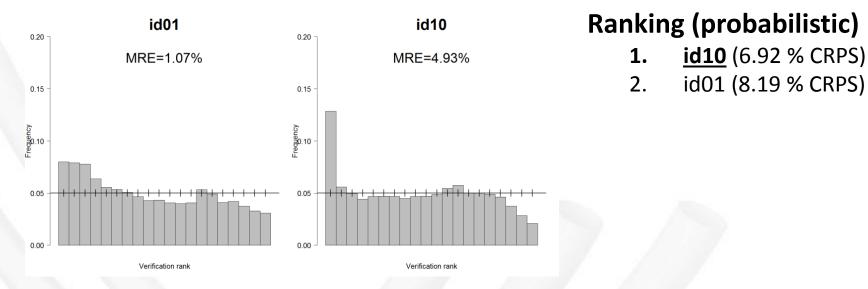
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Klim results (probabilistic)



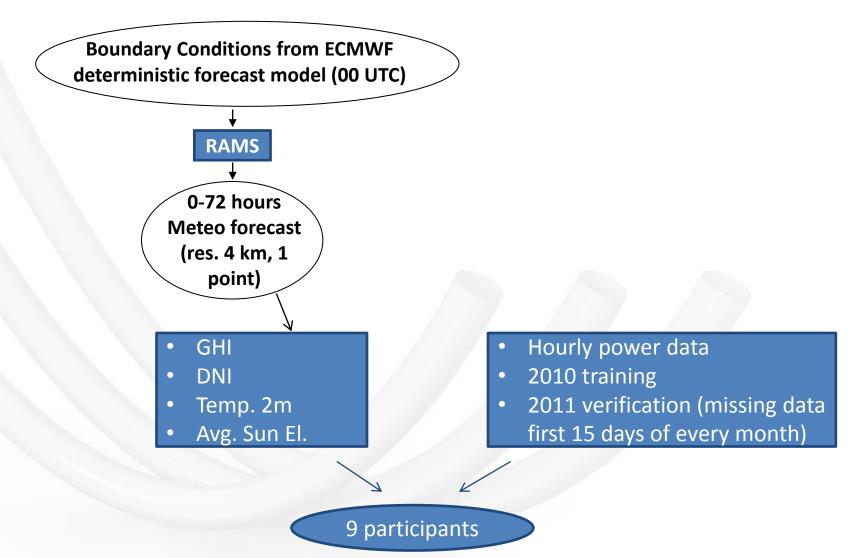
Rank histogram

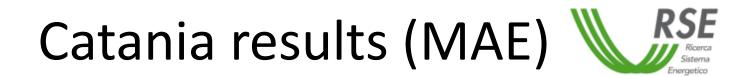


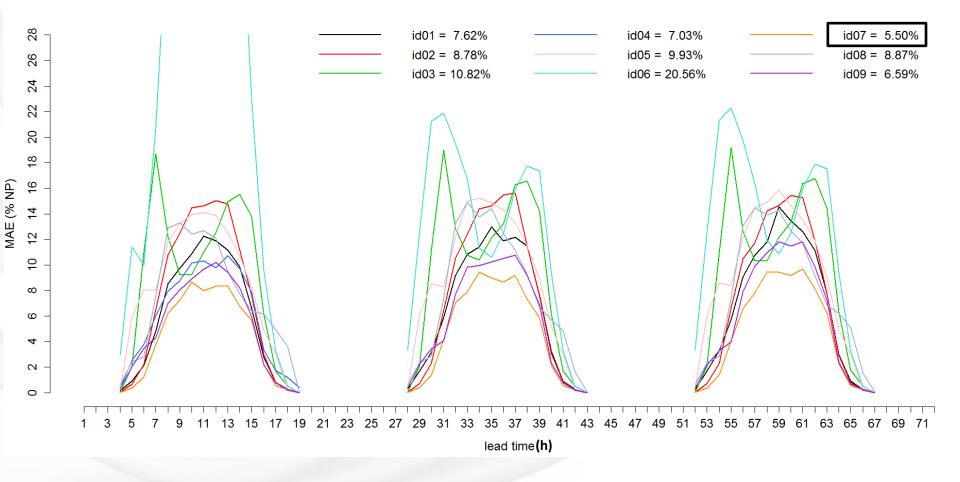
- Conditional kernel density estimation with a quantile-copula estimator.
- Inputs: forecasted wind speed and direction (level 30), hour of the day and leadtime. 5% and 95% quantiles were computed from the forecasted PDF using numerical integration.

Catania case (PV)









Catania results (conclusions)

Ranking

- 1. <u>id07</u> (5.50 % MAE)
- 2. id09 (6.59 % MAE)

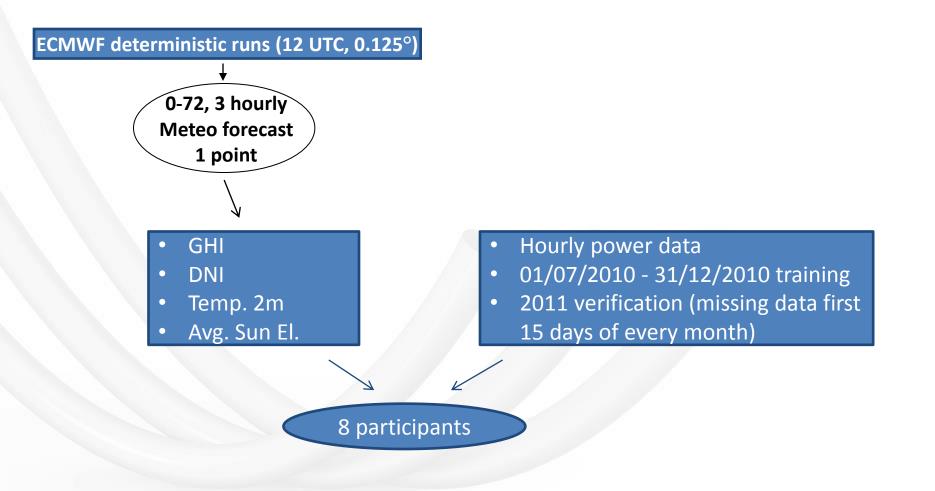
Diebold-Mariano statistic

- H0: the two models have the same forecast accuracy
- H1: 2nd model is less accurate than the 1st one
- p-value=<u>0.998</u>

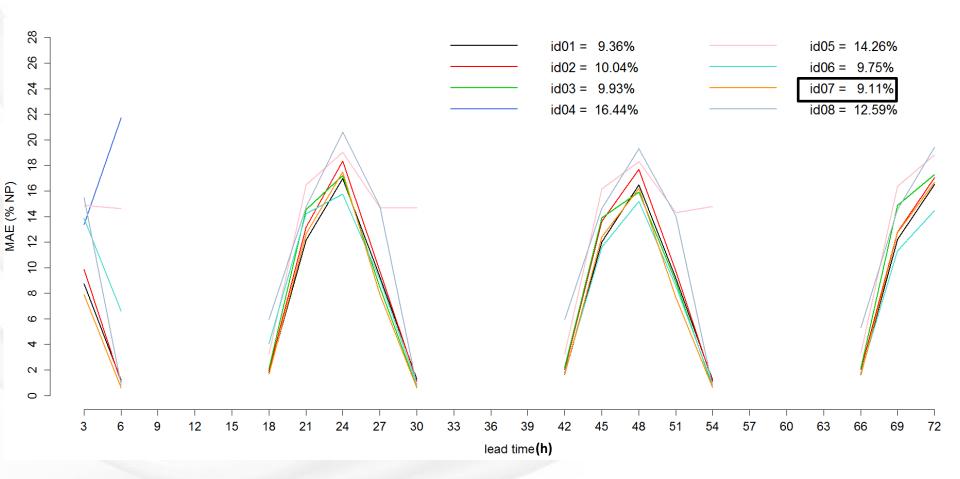
- Meteorological model: RAMS provided by COST
- Quantile regression in order to estimate a clear sky production, a clear sky irradiation and a medium temperature.
- Linear regression to explain the rate of clear sky production observed.
- Bias correction depending on time and the power forecasted using although a quantile regression.

Milano case (PV)









Milano results (conclusions)





Ranking

- 1. id07 (9.11 % MAE)
- id01 (9.36 % MAE) 2.

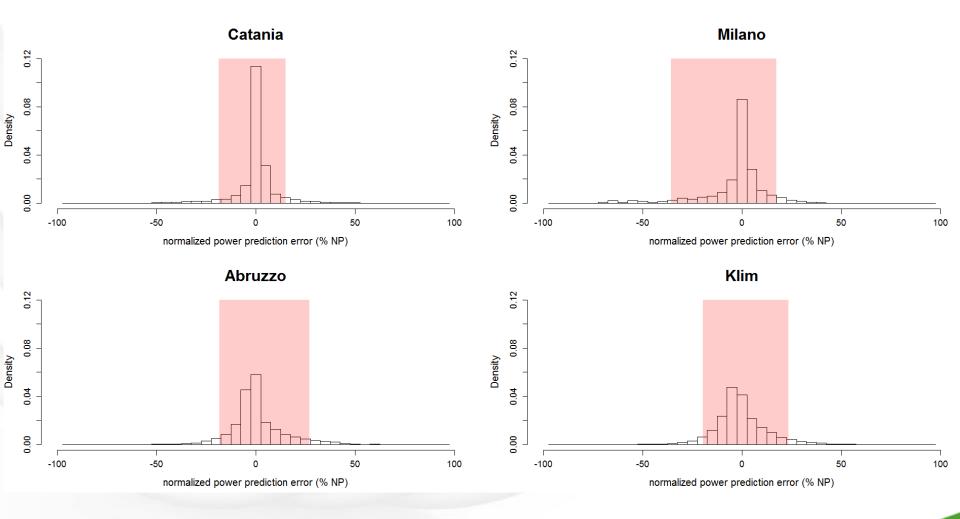
Diebold-Mariano statistic

- H0: the two models have the same forecast accuracy
- H1: 2nd model is less accurate than the 1st one
- p-value=**0.961**

- Meteorological model: RAMS provided by COST
- Quantile regression in order to estimate a clear sky production, a clear sky irradiation and a medium temperature.
- Linear regression to explain the rate of clear sky production observed.
- No bias correction (not enough data)

Solar vs Wind, errors distributions (of best models)





Shaded area: 5%-95% interval

Conclusions



- Wind power prediction, best model performances: complex terrain MAE = 9.04%, flat terrain MAE = 9.45 %
- Using ECMWF global model data (0.125°x 0.125°) of year 2010-2011 (plus post-processing) in a complex site has allowed to obtain the same performance as those obtained by a Limited Area Model of year 2001-2002 in a flat terrain site
- Solar power prediction best model performances: MAE = 5.5% (Catania, less pollution, greater mean irradiation), MAE = 9.11% (more polluted)
- Solar PV energy seems to be more predictable than wind energy with lower MAE and narrower error distribution only for Catania solar farm
- Same winner for both solar plants (EDF, Christophe Chaussin)
- Same winner for both wind farms deterministic (Fraunhofer Institut, Jan Dobschinski), two for probabilistic (Norwegian Meteorological Institute, John Bremnes and INESC, Ricardo Bessa)