

## **Production Loss estimation for wind power forecasting**

Winterwind 2014 Timo Karlsson, Ville Turkia, Tomas Wallenius, Jari Miettinen VTT Technical Research Centre of Finland



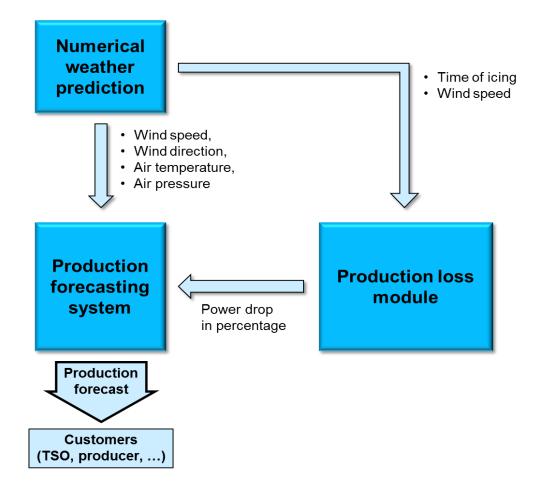
#### **Motivation**

- Icing induced production losses are a known source error for production forecasting
- Need a method to correct the forecasts to include icing events
- Forecasting systems differ
  - Need something that can work in more than one
- Simple solution preferred



## **Module for forecasting**

- estimate production losses caused by icing
- use real production data
- Look at effects of rotor icing
- Quantify the actual production losses



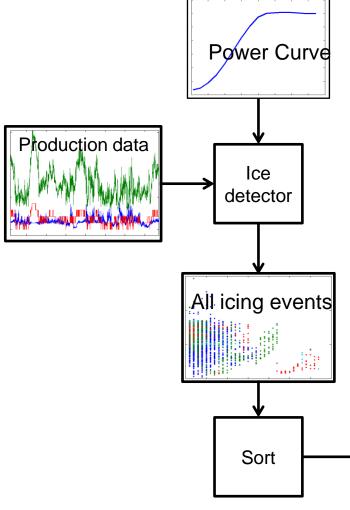


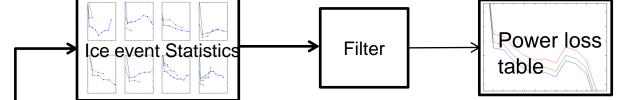
## **Design Goals – KISS principle**

- Portability
  - not dependent on a certain forecasting system
- Generality
  - results not case/site dependent
- Simplicity
  - minimize the amount of needed inputs



## Method





- collect production data from several sites
- 2. Detect icing events
- calculate the production losses during icing events
- 4. calculate collected statistics on these events
- 5. combine the statistics to get a generic result



#### Data

- Two sites in Finland
  - One in North (site A, two turbines)
  - One in South (site B, one turbine)
- different types of turbines
- combine different sites to each other
- Two years of data on both sites
- Production data, no external ice detection



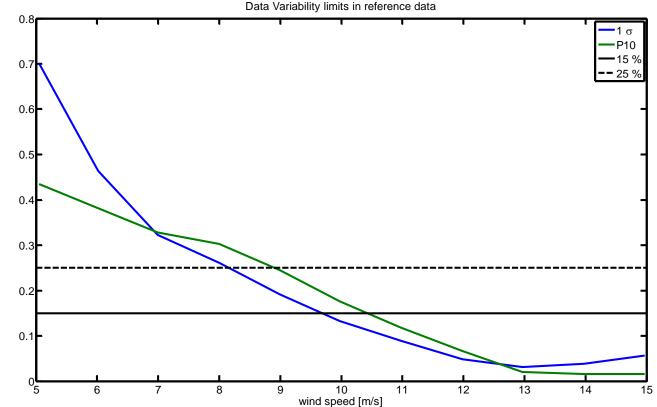
## **Icing event defintion**

- Rotor icing detected from production data
- Often used metric is a constant power drop
  - -15..-25% from reference
- Data variability is larger on lower wind speeds
  - Limit should be wind speed dependent
- Lots of outliers
  - Time dependence
  - Robust statistics
    - P10 rather standard deviation

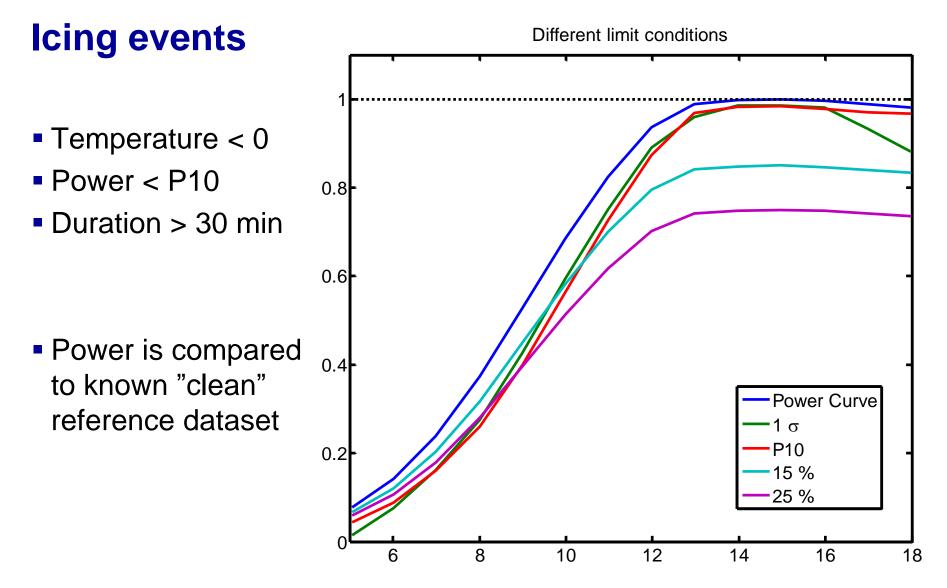


## **Icing event definition**

- Data variability:
  - wind speed dependent
  - Large at low wind speeds
- At low wind speed power 25 % below reference is not out of the ordinary

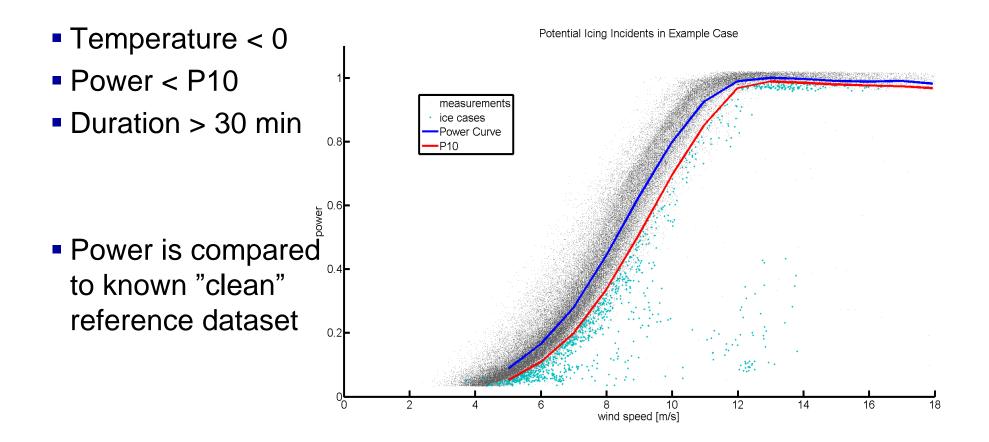








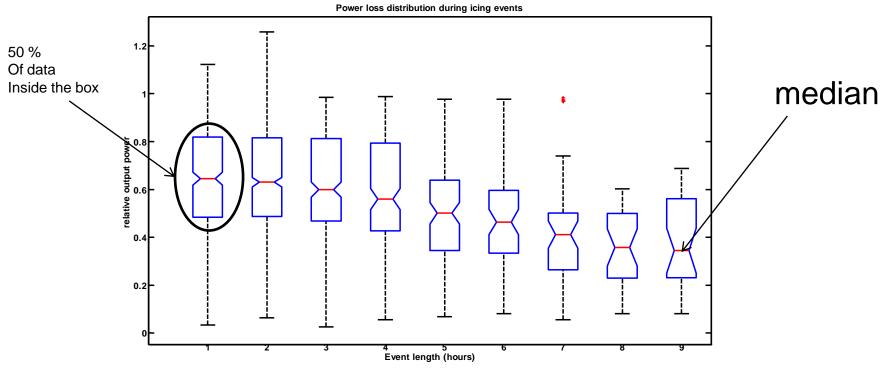
## **Icing events**





#### **Loss statistics**

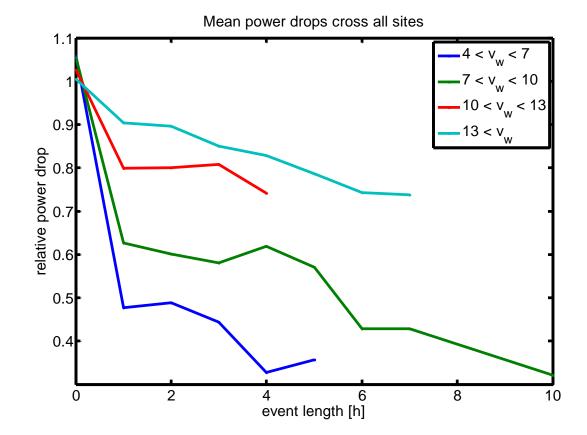
- Massive variance in the data
- big median losses even after just one hour
- median loss trending down along with event length





### **Loss Statistics**

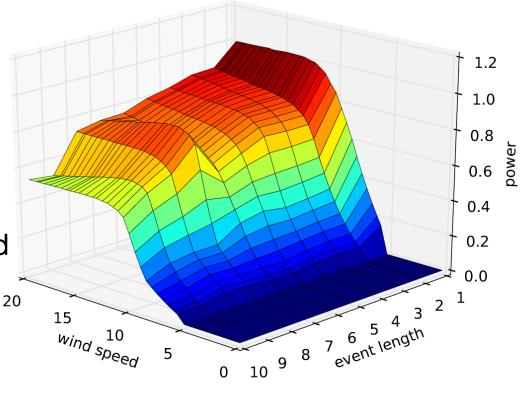
- Differences between sites are visible
  - Caused by differences in data variability
- Icing effect a lot more dramatic on smaller wind speeds





## **Conclusions from statistics**

- Event length matters
  - Longer icing event -> bigger losses
- Wind speed has an effect
  - Relative losses significantly bigger at small wind speeds
- Site by site differences caused by data variability





## **Preliminary look at forecasting**

Resulting method power really sensitive to icing forecast forecast pov icing forecast 9.0 ke bower Relati accuracy 0 Timing another 2011-11-30 00:00:00 2011-12-01 2011-12-02 00:00:00 2011-12-03 00:00:00 2011-12-04 00:00:00 2011-12-05 00:00:00 2011-12-06 00:00:00 2011-12-07 00:00:00 2011-12-08 00:00:00 2011-12-09 00:00:00 source of errors Meteorological 0.25 vs. rotor icing 0.2 Icing indicator 0.05



### **Sources for forecast errors**

- Site-specific differences
  - Large data variability -> larger predicted power drops
- Stops
  - Losses caused by unscheduled, unpredicted stops can be larger than losses caused by decresed production
  - Model needs to be able to predict stops
  - Ice mass, if available
- Timing



## **Conclusions**

- Production losses
  - Can be estimated from past production data
  - Wind speed
  - Icing event definition
  - Stops
- Some similarity accross sites

- Forecasting:
- Error sources
  - Timing
  - Rotor vs. meteorological icing
  - Production loss accuracy



#### **Future work**

- More work on forecasting
  - Different time windows
  - Different kinds of icin forecasts
  - More sites
- Later this year

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