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Forecasting Production Losses at a Swedish Wind Farm

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Forecasting Production Losses at a Swedish Wind Farm

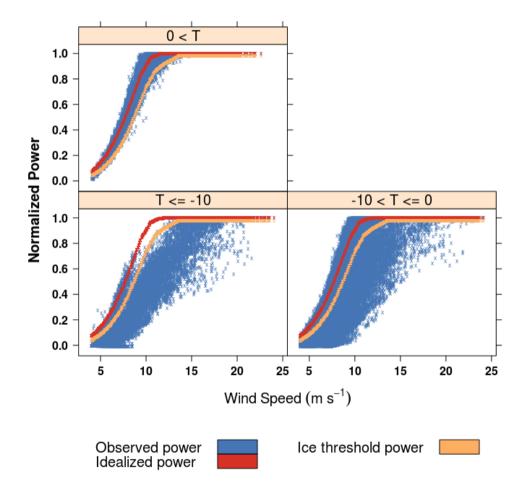
WinterWind 2013 Neil Davis^{1,2}, Andrea Hahmann¹, Niels-Erik Clausen¹, Mark Zagar², and Pierre Pinson³ $f(x+\Delta x)=\sum_{i=0}^{\infty} \frac{(\Delta x)^{i}}{i!} f^{(i)}(x)$ $f(x+\Delta x)=\sum_{i=0}^{\infty} \frac{(\Delta x)^{i}}{i!} f^{(i)}(x)$

1: DTU Wind Energy; 2: Vestas Wind Systems; 3: DTU Informatics

DTU Wind Energy Department of Wind Energy



- Site location
- Wind park planning
- Energy market pricing

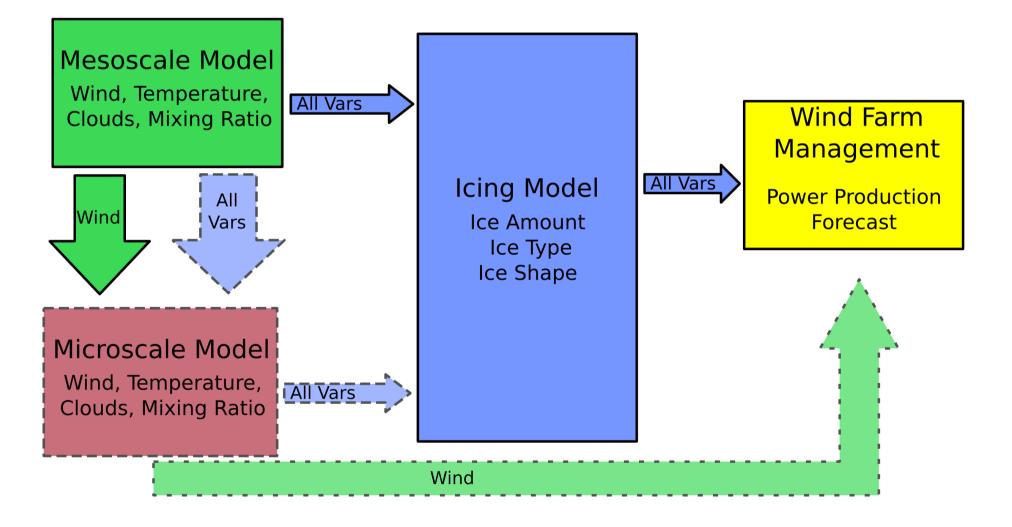


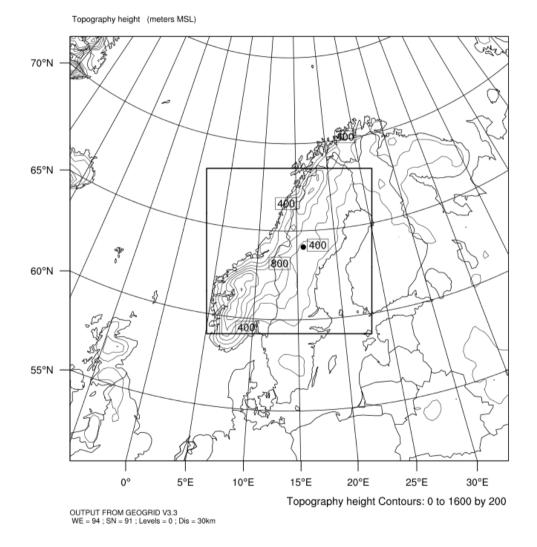
Wind Park Power Curve





Production Forecast Model



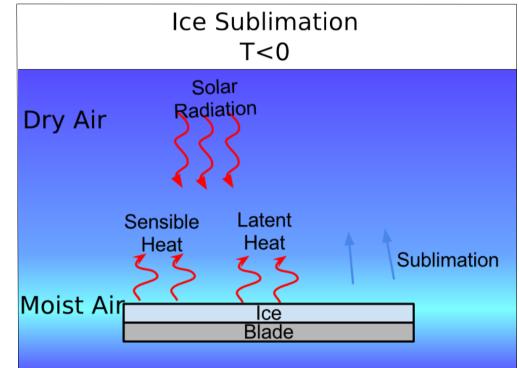


- Observational data
 - · Located in central Sweden
 - Approximately 50 Vestas V90 turbines
 - · Grouped into 3 parks
 - Observations from January 2011
 - Temperature, wind speed , wind direction from hubs of each turbine, production & turbine normal operation time
- WRF mesoscale simulation
 - 27 km & 9 km nests
 - Thompson microphysics & MYNN2 PBL
 - Best performing of 9 sensitivities
 - FNL for initial & boundary conditions
 - Grid nudging on the outermost domain
 - 63 vertical levels



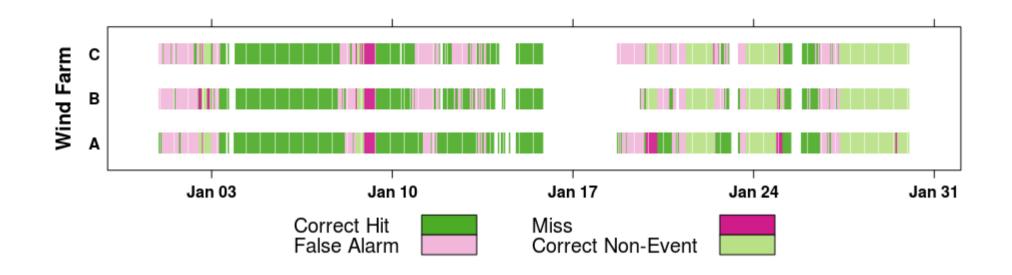
Icing model

- Modified Makkonen model
 - Cylinder moves at blade relative velocity
 - Diameter 0.144 m
 - Located at 75% of blade length
 - · Heat transfer coefficient for airfoils
 - Blade always at 80m hub height
 - Utilize all 4 WRF hydro-meteor types (QCLOUD, QRAIN, QICE, QSNOW)
- Sublimation & shedding included
 - Sublimation based on humidity gradient & radiation balance
 - Shedding set to 100% when T > 1° Celsius





Ice duration evaluation



- Timeseries comparing model icing periods to periods when any turbine was iced in a given farm.
- Compared with persistence & threshold method for several skill scores and this method outperformed both
- For more details see paper submitted to Journal of Applied Meteorology & Climatology "Forecast of Icing Events at a Wind Farm in Sweden"

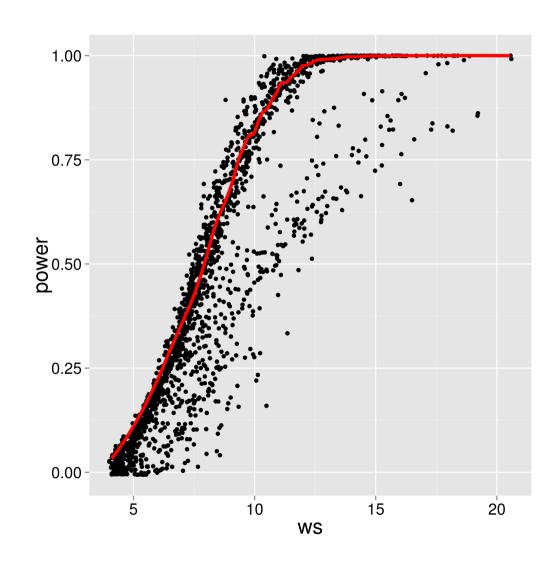
DTU Wind Energy, Technical University of Denmark

Production loss model

- Fit smoothing function to power curve
 - Wind farm average values
 - Only for temps above freezing
 - Red line in the plot

Vestas

- Calculate power difference
 - Deviation from modeled power
- Investigate potential predictors for power difference
 - Ice Model outputs
 - WRF Hydrometeors
- Fit test models for all farms
 - Make use of entire dataset
 - Maximize adjusted R²





Model Parameters

• Threshold

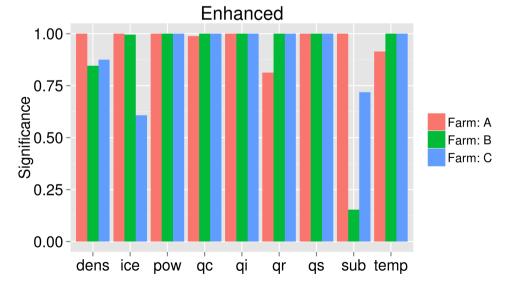
Vestas

- Model qall > 1e-3
- Set power to 0
- Use power curve all other times

- Ice only
 - Forecasted power
 - Accumulated mass
 - Average ice density
 - Sublimation
 - Temperature

- Enhanced
 - Ice only parameters
 - Square root of all 4
 hydrometeors

February 21, 2013

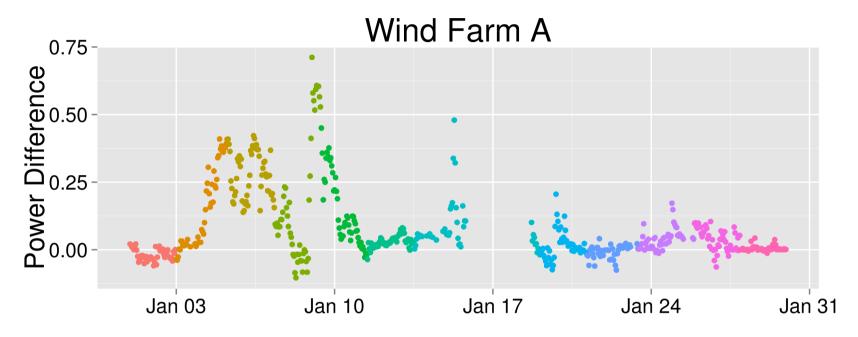








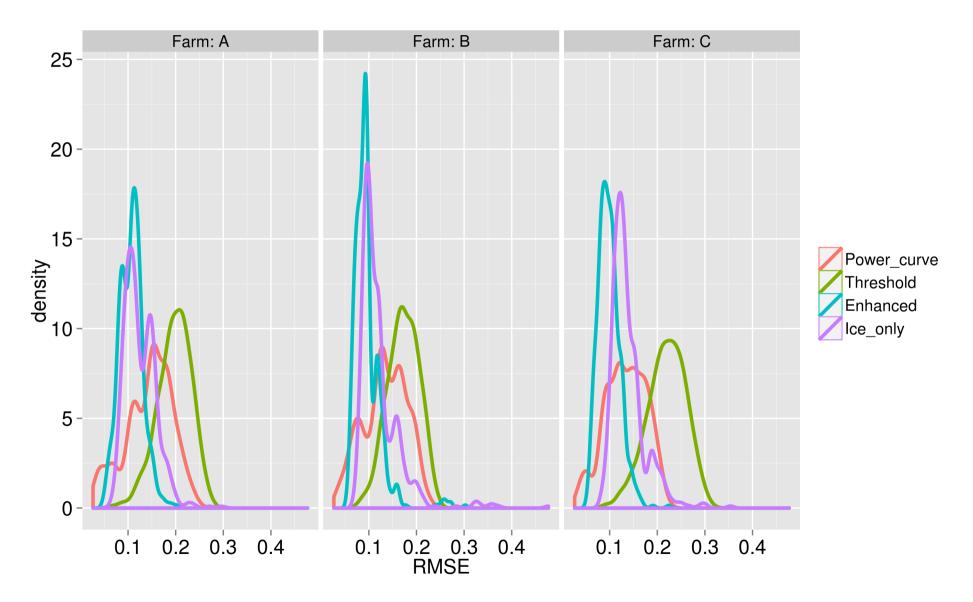
K-fold cross validation



- Cut into 12 pieces
- Fit 8 pieces (training), and forecast remaining 4 (test)
- Calculate RMSE & mean bias of mean farm power forecast (test)
- Monte Carlo approach with 495 different model fits

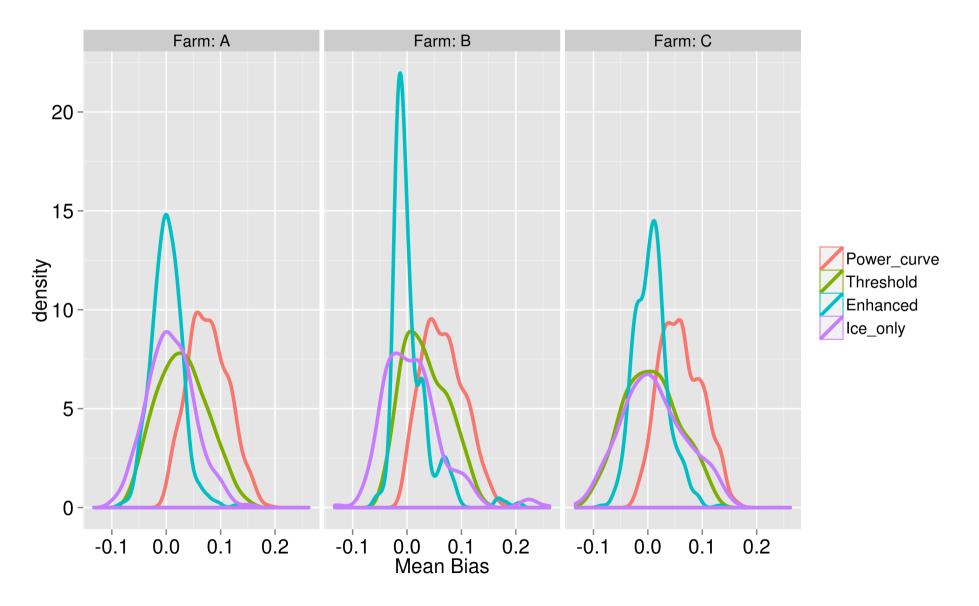


RMSE











Conclusions

- Combination of WRF output parameters & icing model parameters works best for all 3 wind parks
- Both bias and RMSE of hourly production estimates can be improved using this approach
- Both statistical approaches show improvement over the threshold based method
- For this site the icing model output was a secondary feature, with the cloud outputs from WRF performing as well as the icing model.
 - We propose this is due in part to the very cold temperatures during icing, so the physical icing model does not have as much impact.

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Future Work

- Apply this method to other sites and longer periods
 - Investigate possible time lags using time series analysis
- Ensure the modified Makkonen model is representing the turbine icing correctly
 - Develop relationships between the two if required
- Enhance the formulation of ice removal mechanisms
- Evaluate performance using forecasted winds



Questions???