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Turbine-specific pad-level ice loss assessment

Advantages, accuracy, and challenges

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Value drivers for improving ice prediction

- Cold climate presents about 10% of the market.
- Production loss due to icing has large impact on business case (cost of energy).
- Active mitigation comes with a price.
- So far the total park production loss has been estimated.
- Knowledge about intra-park variability presents opportunities.

*Ice Assessment** enables more precise and accurate business case analysis

• Large differences in conditions exists within a park.

Example:

 Determine if IPS (Ice Protection System) is needed for all turbines in a park and by this reduce cost

*Product name for turbine specific pad-level ice loss estimate

Production loss

What we observe and attempt to predict for the lifetime

- Production data analysis
- Task19 or equivalent approach
- Park, Average

	Ice Loss (%AEP)
Subtotal	9,1 %
2010-2011	7,0 %
2011-2012	5,0 %
2012-2013	14,0 %
2013-2014	13,5 %
2014-2015	9,3 %
2015-2016	6,3 %

- Variability by years
- Intra-park variability (individual turbines)

Observations from one wind farm

Total park production loss



Production loss separated by turbines



Ice Assessment methodology

Process and techology



Technology 1: Microscale weather modelling (WRF-LES)

Lack of ice-specific observation; complemented with validated numerical weather models

Modelled vs. Observed

(somewhere in Scotland)

Or maybe Modelled vs. Observed



Technology 1: Microscale weather modelling (WRF-LES)

Increasing precision

- More accurate terrain
- Additional physical and weather features



What to see in the plots below?

- Higher terrain more ice accretion
- Production losses vary a lot between WTG's
 - Size of blue circles proportional to loss

Accr (kg)



Technology 2: iceBlade ice accretion model (Davis et al. 2014)

 Rate of ice mass growth on a rotating cylinder (not designed for wind energy; Makkonen 2000)

$$\frac{dM}{dt} = \alpha_1 \alpha_2 \alpha_3 \cdot w \cdot A \cdot V$$

- Empirical equations (Scaled 0 to 1):
 - Collision efficiency (α_1)
 - Sticking efficiency (α_2)
 - Freezing efficiency (α_3)



Homola, M. C., T. Wallenius, L. Makkonen, P. J. Nicklasson, and P. A. Sundsbø, 2010: The relationship between chord length and rime icing on wind turbines. Wind Energy, 13, 627–632, doi:10.1002/we. Enhanced for wind turbines (Davis et al; 2014)



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7 Turbine-specific pad-level ice loss assessment (Public)

Technology 2: iceBlade ice ablation and accumulation model (Davis et al. 2014)

- Ablation
 - Total shedding when temp above 0 for 30 minutes
 - Sublimation from Thompson microphysics
 - Empirical wind erosion
- Accumulation
 - Combine Accretion and Ablation
 - Limit max ice amount



Technology 3: Production loss calculation

Multi-parametric power loss curve derived statistically combining:

- Observed performance degradation of turbines when ice is present on the blades
- Predicted ice-related conditions on future projects where performance data is not yet available
- Different ice-affected power curve for every Vestas turbine type



⁹ Turbine-specific pad-level ice loss assessment (Public)

Technology 3: Production loss calculation

"Loss surfaces"



Based on a generalized linear model and training dataset of a number of wind farms and winter seasons

Validation

We want to predict absolute park losses and relative intra-park variability



What causes the intra-park variability

What we can see in the data and what we see but not understand

- Height explains about 60%
- · Ice rose interpretation desired









Future work, Challenges





Comprehensive analysis of factors affecting the remaining errors

What predictors should be added into the statistical model?



y = 0.9583x

Developable platform

Lessons will be learned through future use

- Enhancing weather model capabilities regarding icing conditions
 - It is a specific problem requiring dedicated research
 - Data needed for fundamental understanding and model improvement
 - Systematic (like AI analysis of camera pictures)
 - Relevant quantities: LWC, droplet size distribution, cloud base height, visibility
 - Individual components of the chain (wind,icing) → (power,loss) must be further validated
- Further improvement of the learning: ice loss models learn from direct ice observations, not just power loss
 - Statistically
 - Physically
 - Iced blade aerodynamics, 3D CFD
 - Understanding ablation



Performance of 4 different ice loss models. All are driven by the same weather model input (Davis, 2014)

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