

# Validation of New Model for Short-term Forecasting of Turbine Icing

Using SCADA data from Scandinavian wind farms

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Winterwind 2015

# Contents

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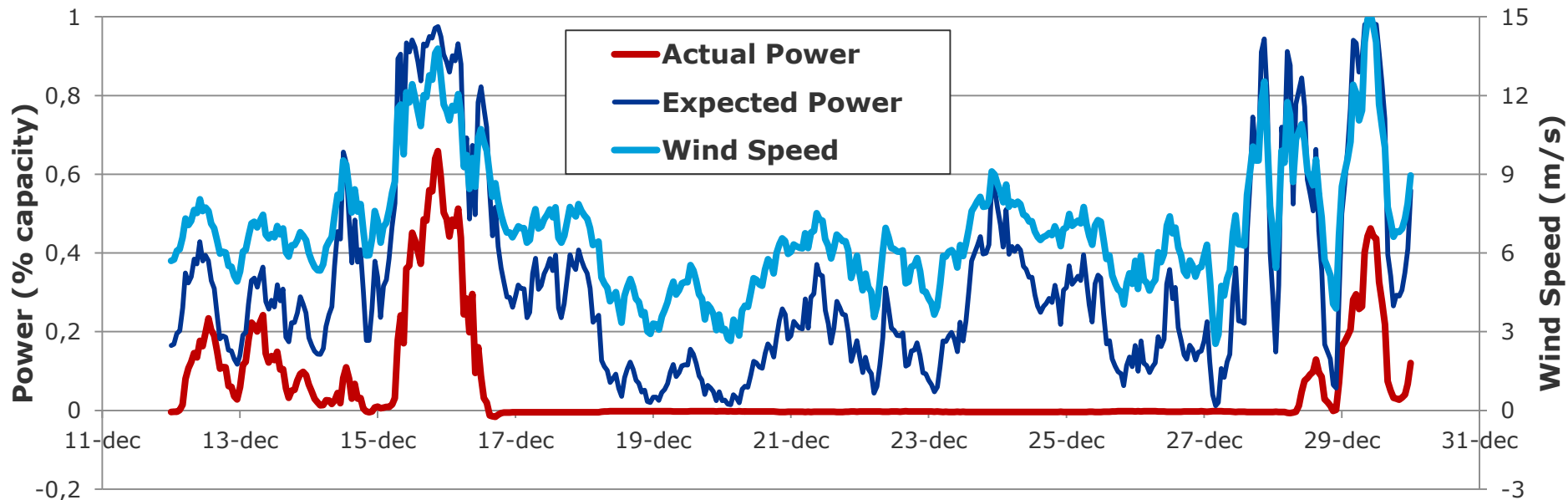
- Background
- Methods
- Validation data
- Results
- Value to forecast users
- Conclusions

## Background – Icing losses

- Icing losses in Scandinavia are variable and can be highly significant
  - Annual energy production losses from  $\sim 0\%$  up to  $>10\%$
  - Monthly energy production losses from  $1\%$  up to  $>50\%$

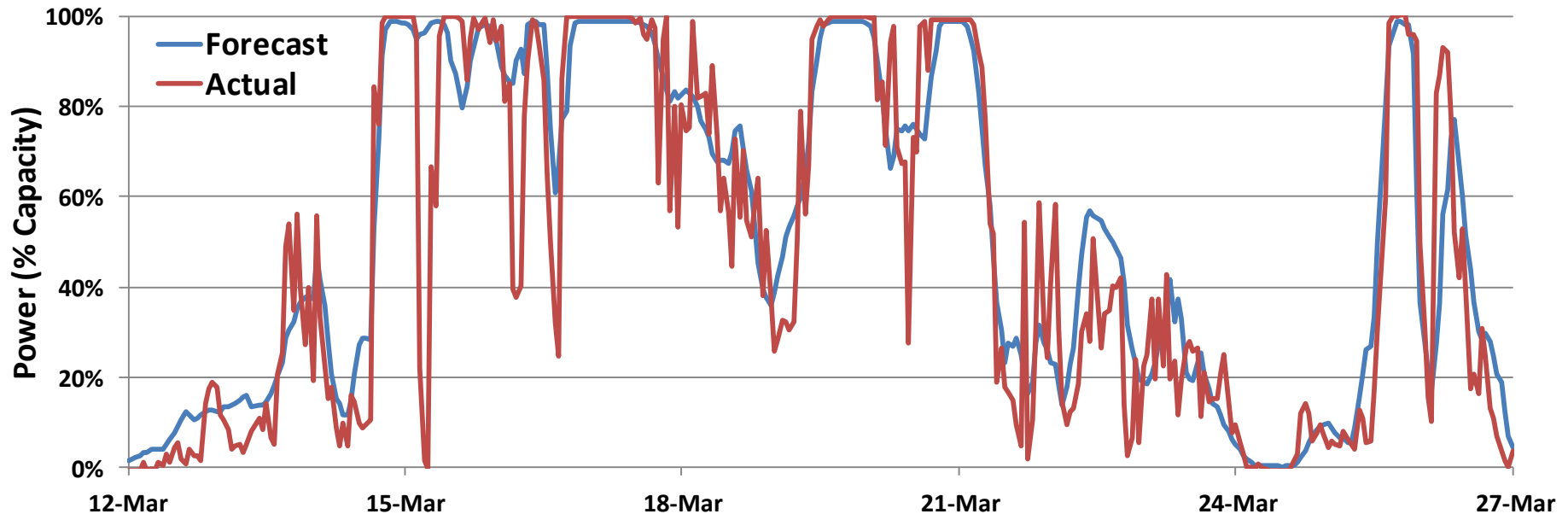
*(Staffan Lindahl: Quantification of energy losses cause by blade icing using SCADA data, Winterwind 2014)*

- Individual icing events can lead to full loss of power

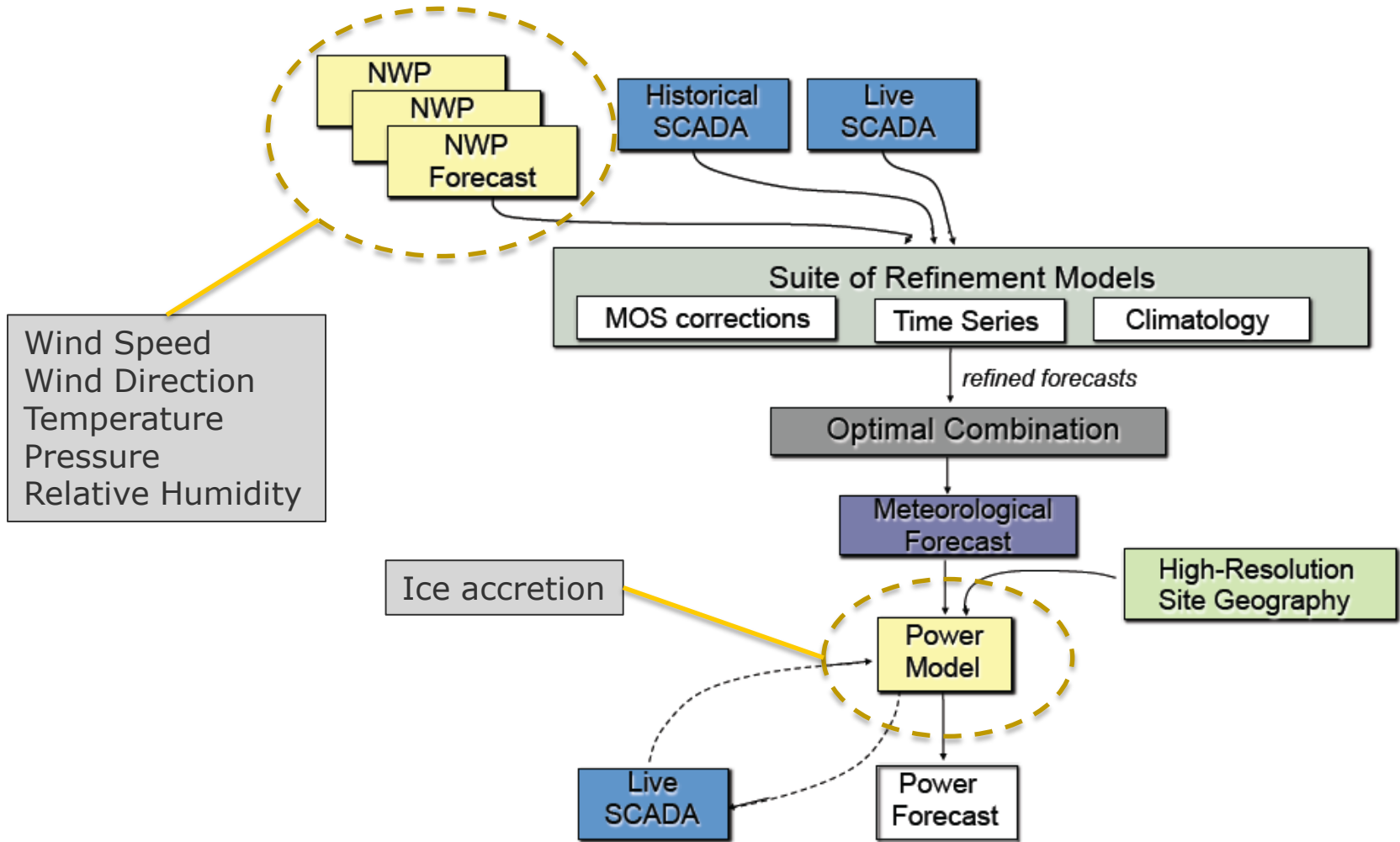


## Background – Short-term forecasting

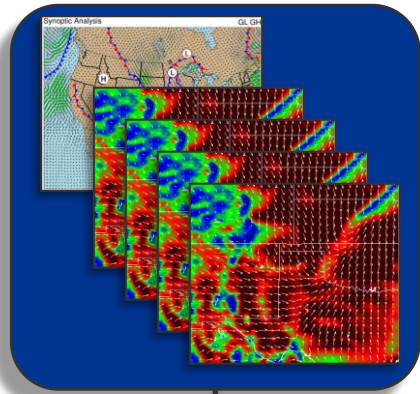
- The value of short-term forecasting is well understood
- State of the art forecasts are typically high accuracy
  - Beneficial to model blade icing when forecasting for wind farms in cold climates
- Icing prediction is woefully unvalidated
  - Reliable observation data is scarce



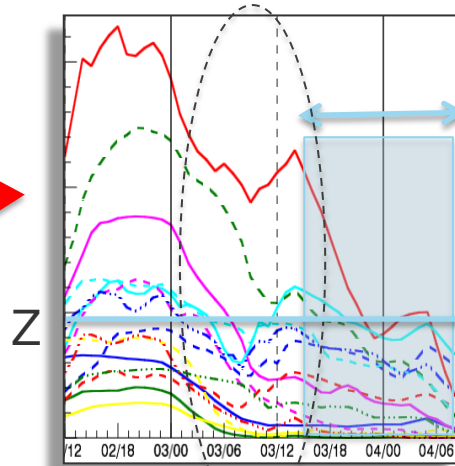
# Methods – base forecast



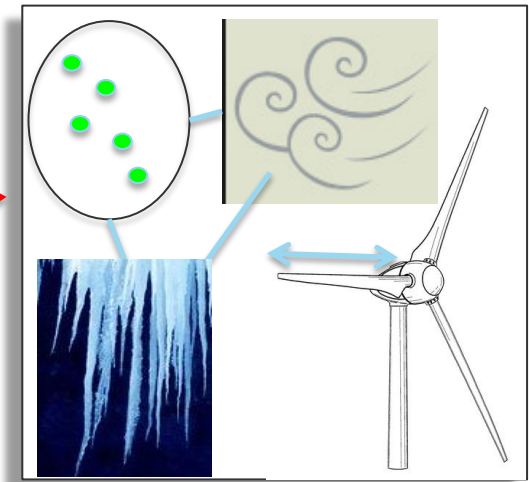
# Methods – icing model



Wind Speed  
Wind Direction  
Temperature  
Pressure  
Relative Humidity

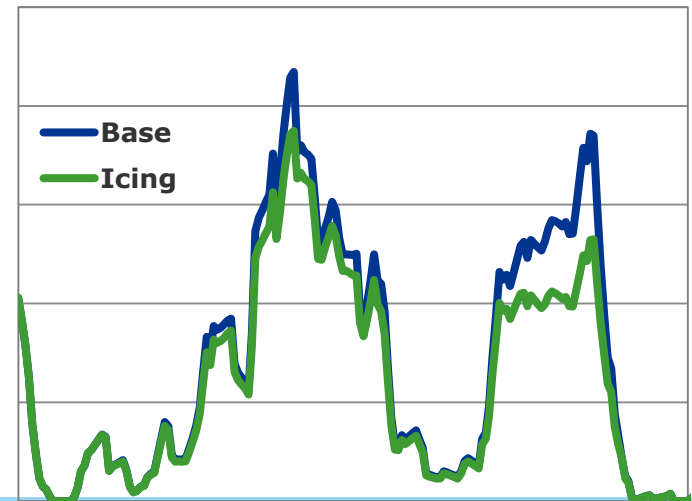


Ensemble Predictions

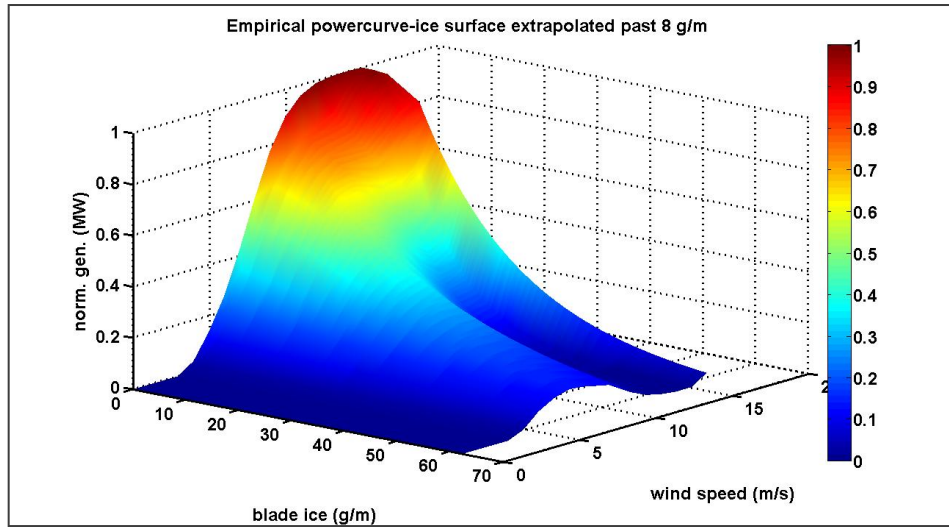


Adapted from Frohboese and Anders (2007)

- Ensemble NWP predictions
- Predicted freezing/thawing time
- Ice accretion parameterization
- Power adjustment



# Methods – icing model power conversion



Borrowed from [http://www.tuuliatlas.fi/icingatlas/icingatlas\\_6.html](http://www.tuuliatlas.fi/icingatlas/icingatlas_6.html)

- Finnish Wind Energy Atlas  
Ljungberg & Niemelä Model
- Converts blade ice, wind speed to power degradation

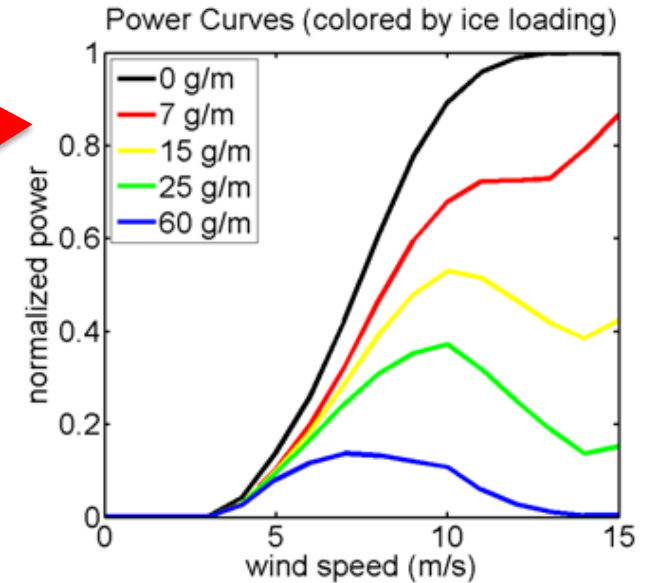
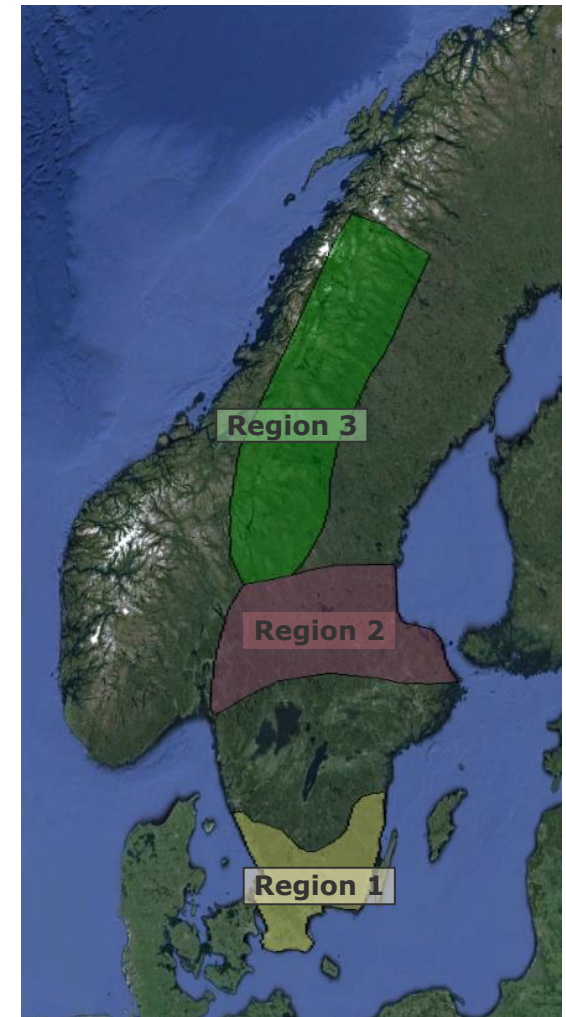


Figure 1: Adapted from Ljungberg, K. and S Niemelä (2011)

## Validation Data

- 3 wind farms, ~40 wind turbines
- Projects in Region 2 and Region 3, where there is sufficient icing to test model
- For each site:
  - ~1 year of data for model training
  - ~1 year of data for validation
- For all projects the turbines remain operational during blade icing periods



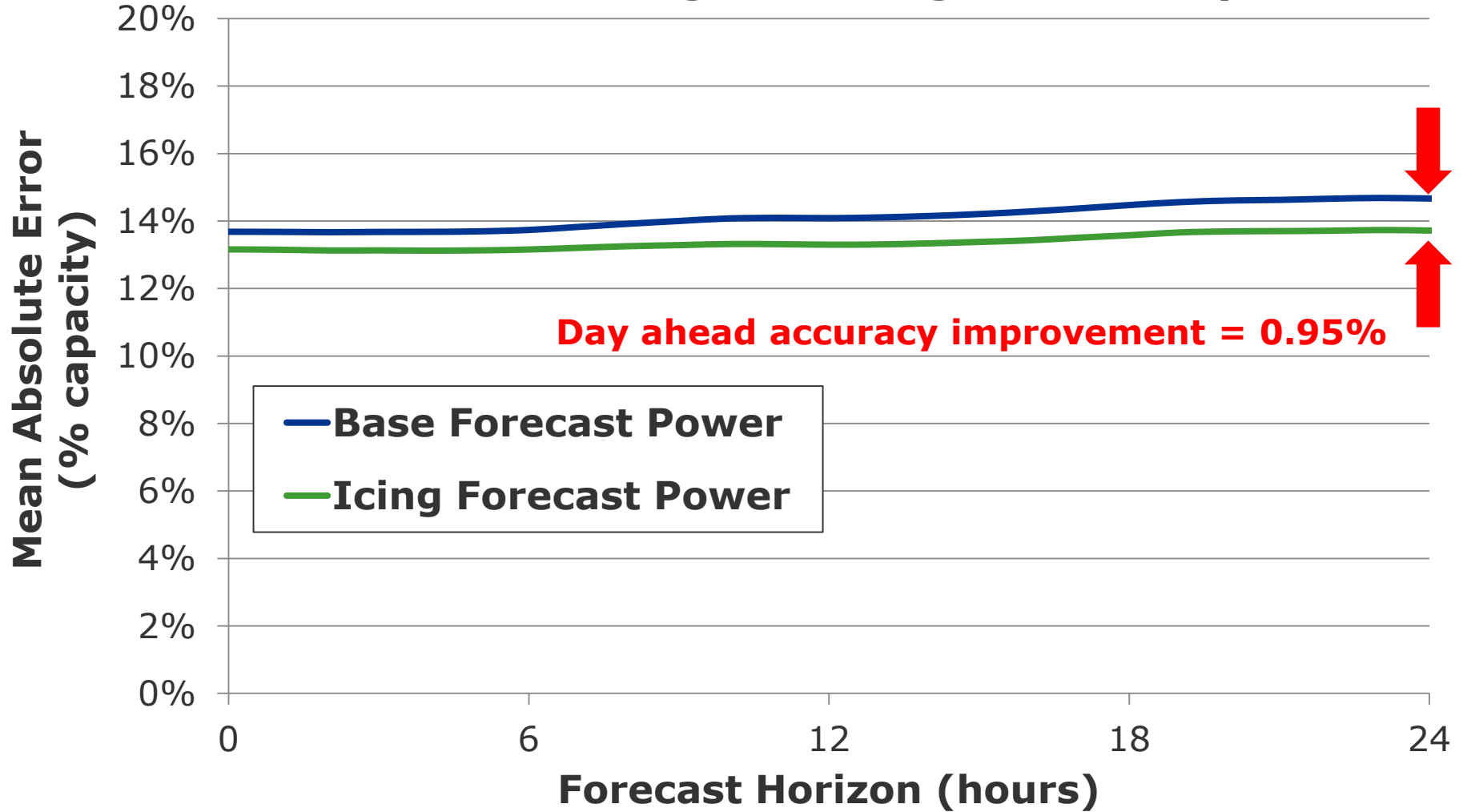


# Results

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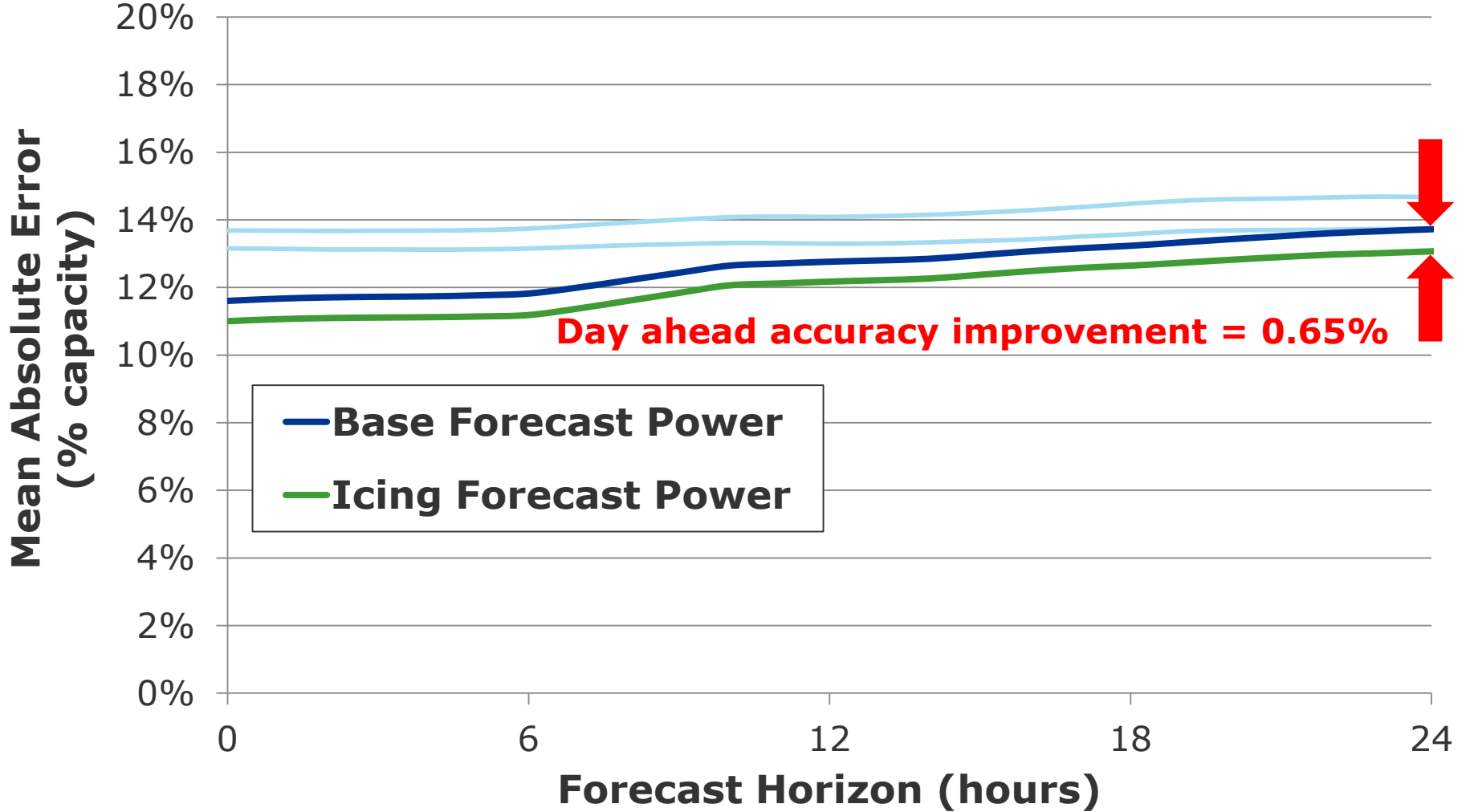
# Results

## Wind Farm 1 – Avg. 13% icing loss over 4 years



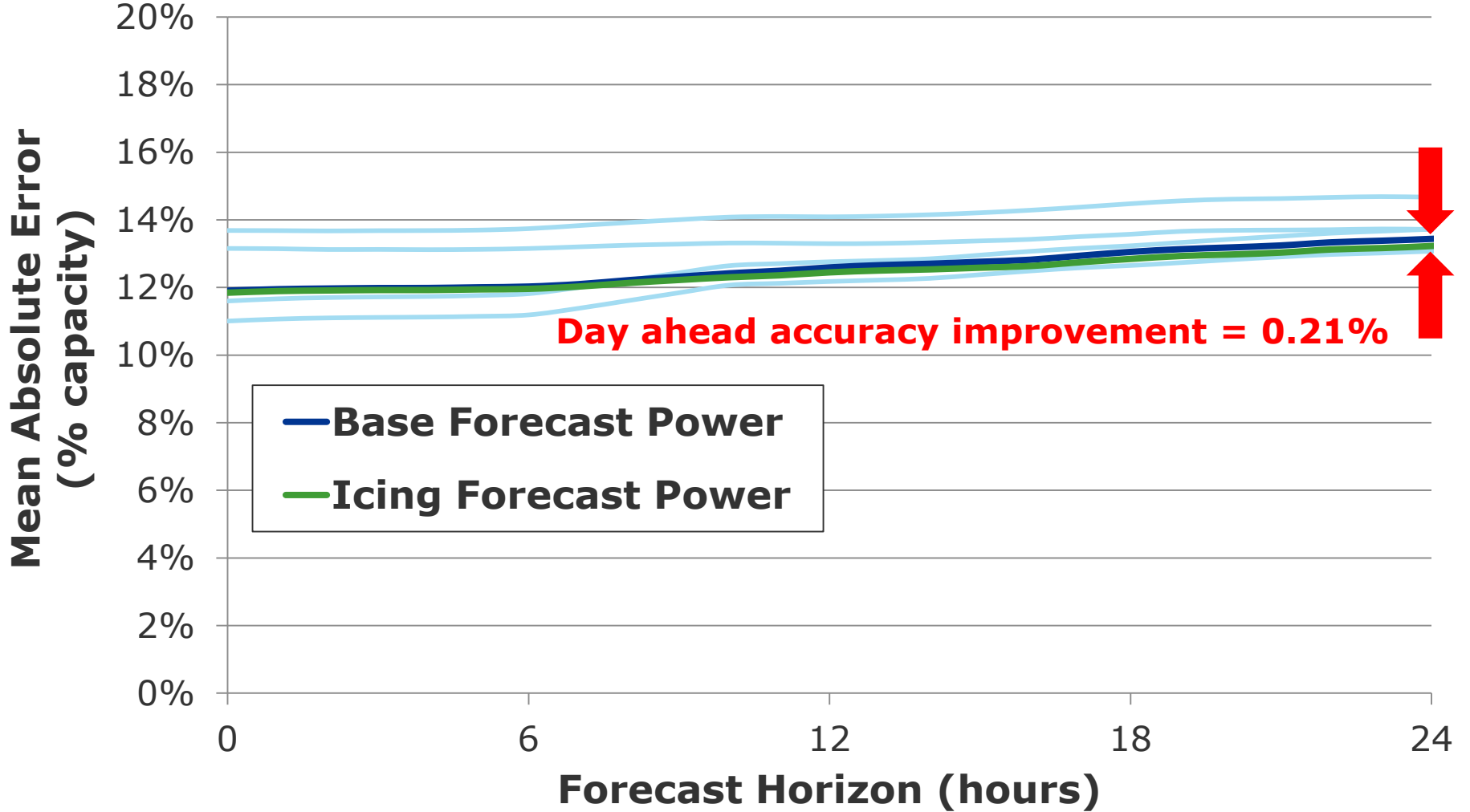
# Results

## Wind Farm 2 - Avg. 6% icing loss over 3.5 years

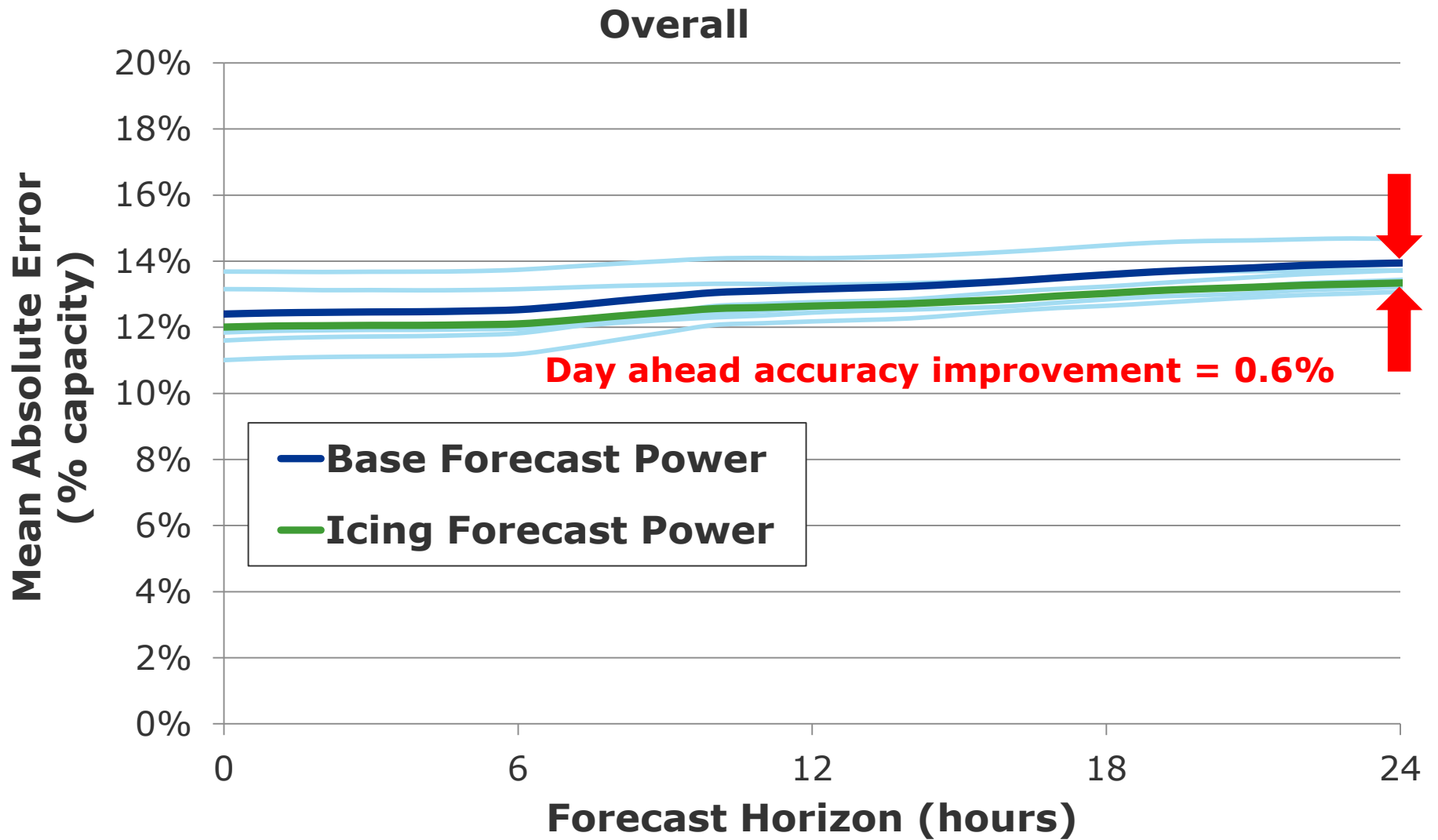


# Results

## Wind Farm 3 - Avg. 4% icing loss over 3.5 years



# Results



## Value to forecast users

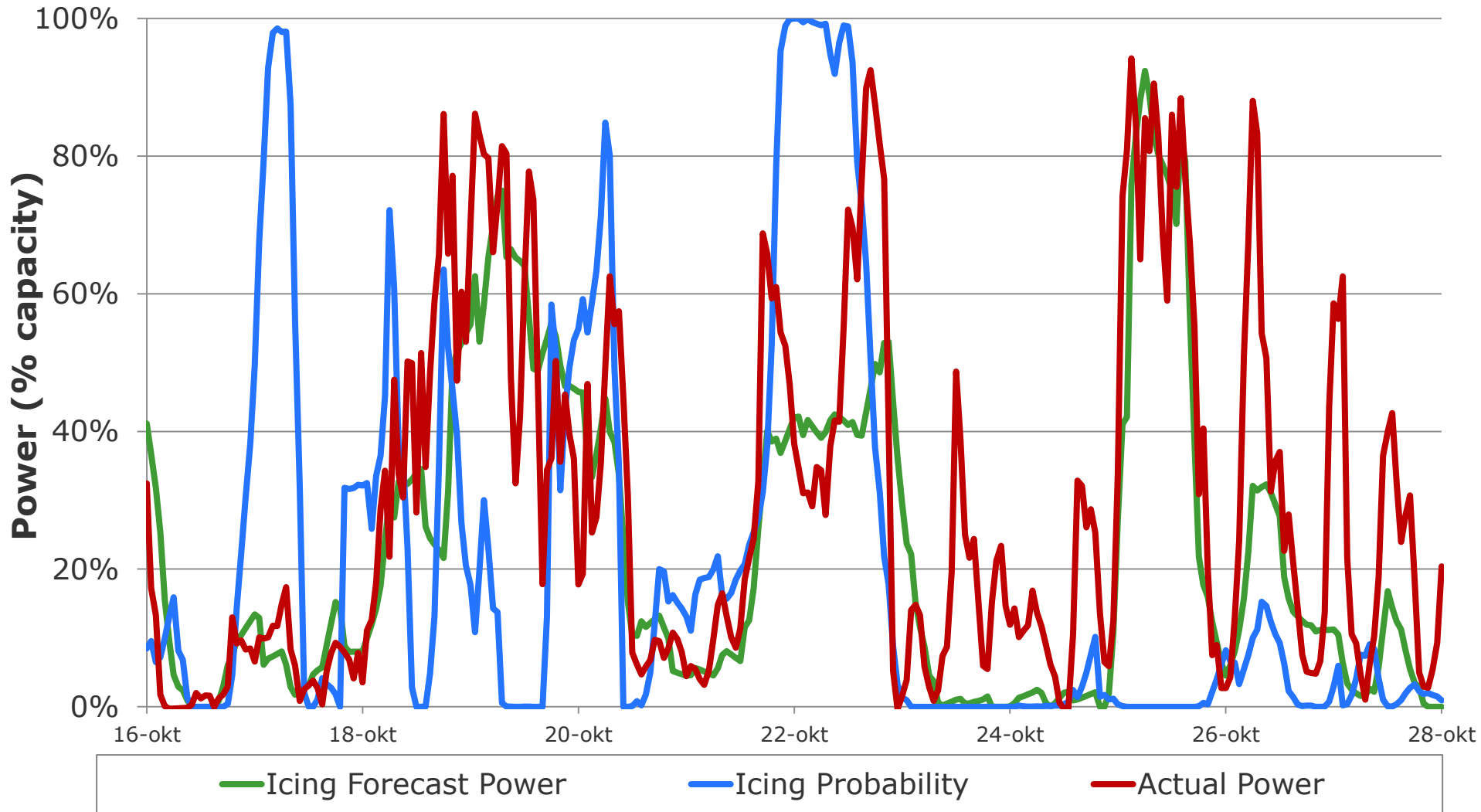
- Icing modelling → improved forecast accuracy
  - Increase energy revenue (based on day ahead energy trading in UK)
  - Operational planning
  - Grid management

Forecast scenario	Average trading revenue (€/MWh)	MAE
No forecast	56.3	-
Basic forecast	57.7	22%
State of the art	61.6	12%
Perfect	64.6	0%

Based on day ahead energy trading in the UK

*Parkes et al. Wind Energy Trading Benefits Through Short Term Forecasting, EWEC 2006*

# Value to forecast users – an advanced warning system



# Conclusions

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- Validation shows icing model adds value to forecasts
  - Model is relevant to Scandinavian climate
  - Successful in varying levels of icing
  - Reduces MAE by up to 0.95% capacity (average improvement = 0.6%)
- Ice accretion → ice load → power is well modelled
- Scope for model improvement
  - Meteorological conditions → icing → freezing time
  - Thawing/ice throw
  - Upper limit for ice load
- Forecast accuracy improvement = increased revenue, informed operations, improved grid management



# Questions?

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