

# ► Validation of day-ahead icing loss forecasts with SCADA data

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# Increasing need for day-ahead icing forecasts



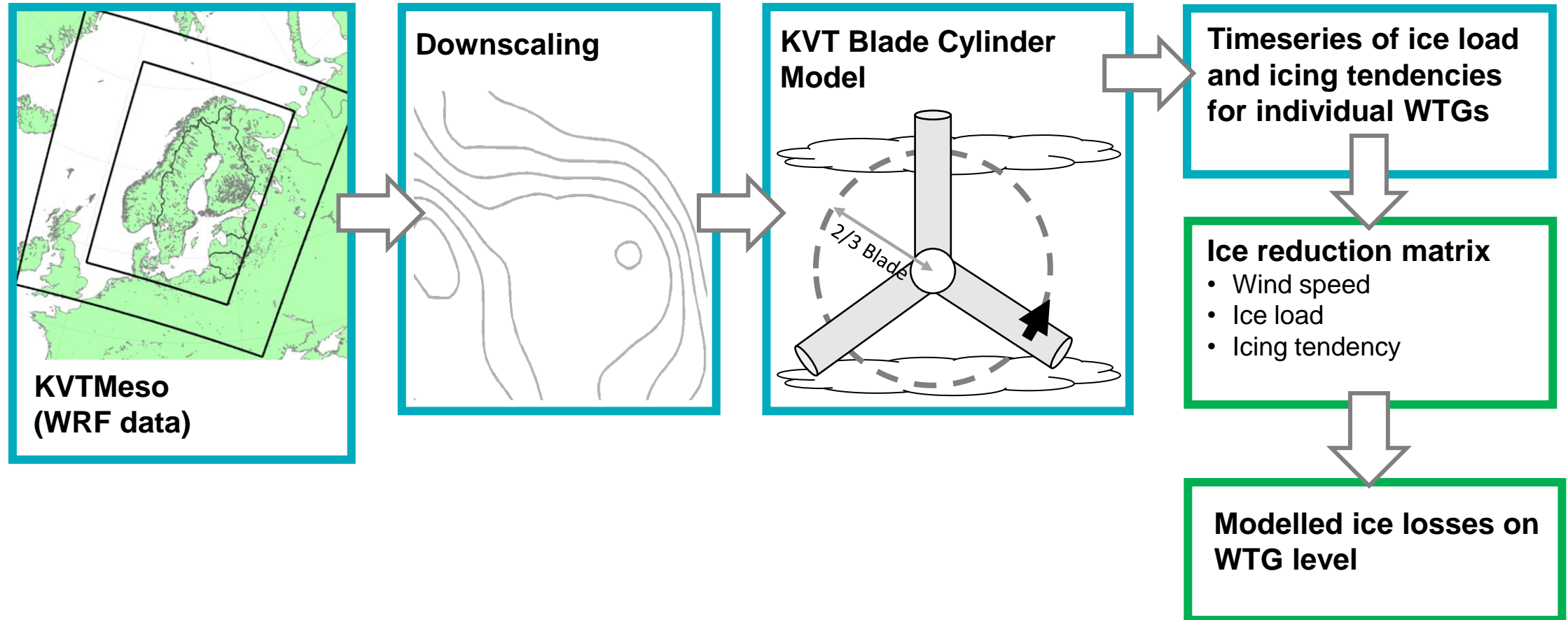
- ▶ The share of wind farms in electricity markets is increasing
  - The impacts of wind power on electricity prices and system dynamics is growing
- ▶ In the Nordics, icing losses are the second highest source of losses after wake losses
- ▶ Bigger turbines → More in-cloud icing in winter
- ▶ Capability to model icing loss for the next day, especially stops, would be crucial for the producer
  - Day-ahead icing forecasts needed for electricity marketing pricing

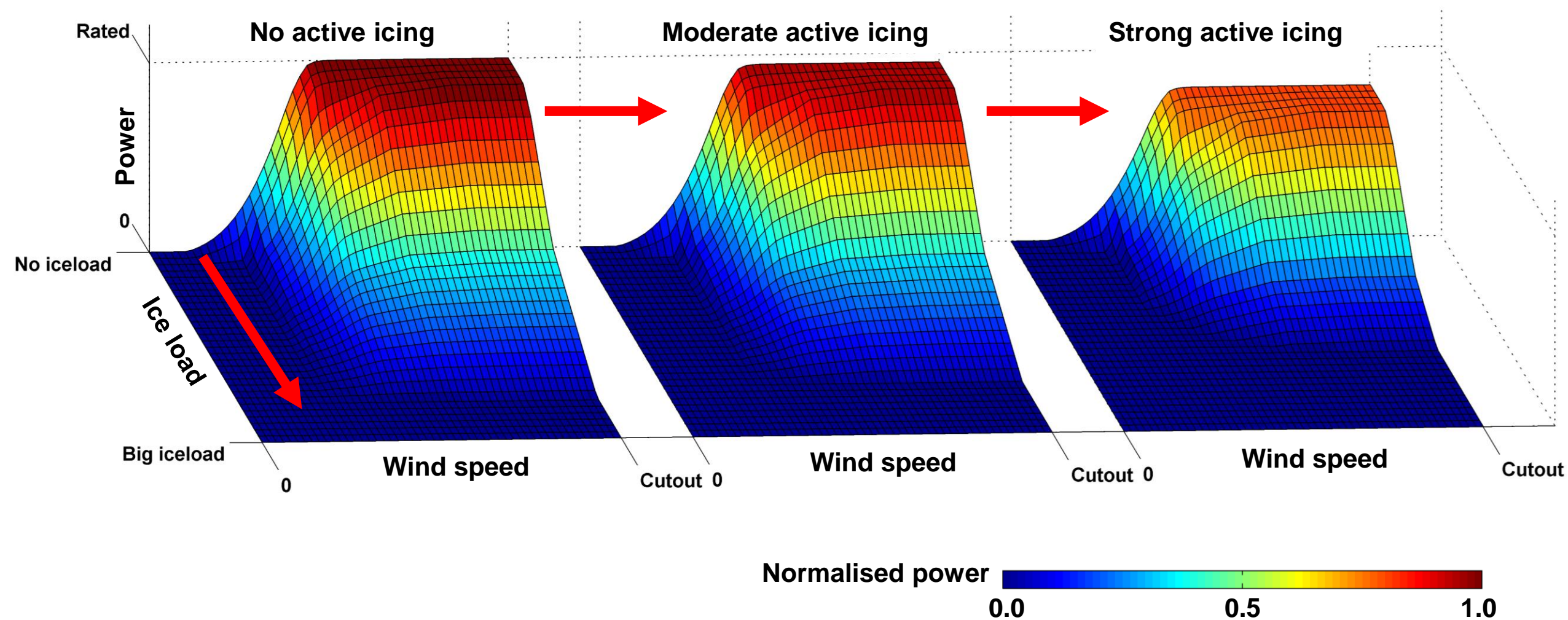
# Operational IceLossForecast 1.0 is being developed from the original IceLoss 2

- ▶ IceLoss 2.1 by Kjeller Vindteknikk
  - ▶ In-house model developed for years
  - ▶ Based on mesoscale weather model WRF combined with in-house post-processing scripts
  - ▶ Used extensively in the Nordics to assess the long-term average icing losses in pre-construction phase
- ▶ Operational IceLoss 2 for short-term forecasts (0-72 h) currently running as a prototype (IceLossForecast 1.0)
  - ▶ Preliminary probabilistic forecasts also included
  - ▶ Further development ongoing

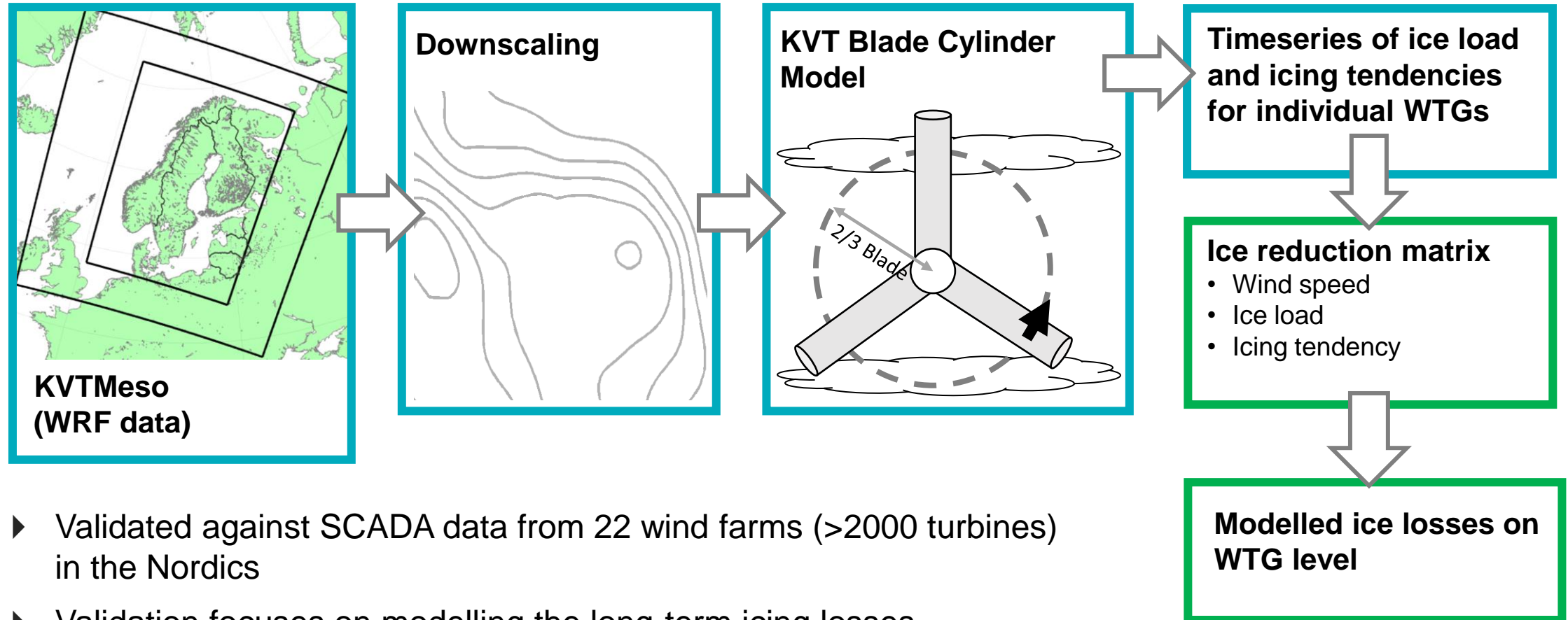


# The IceLoss 2.1 model chain



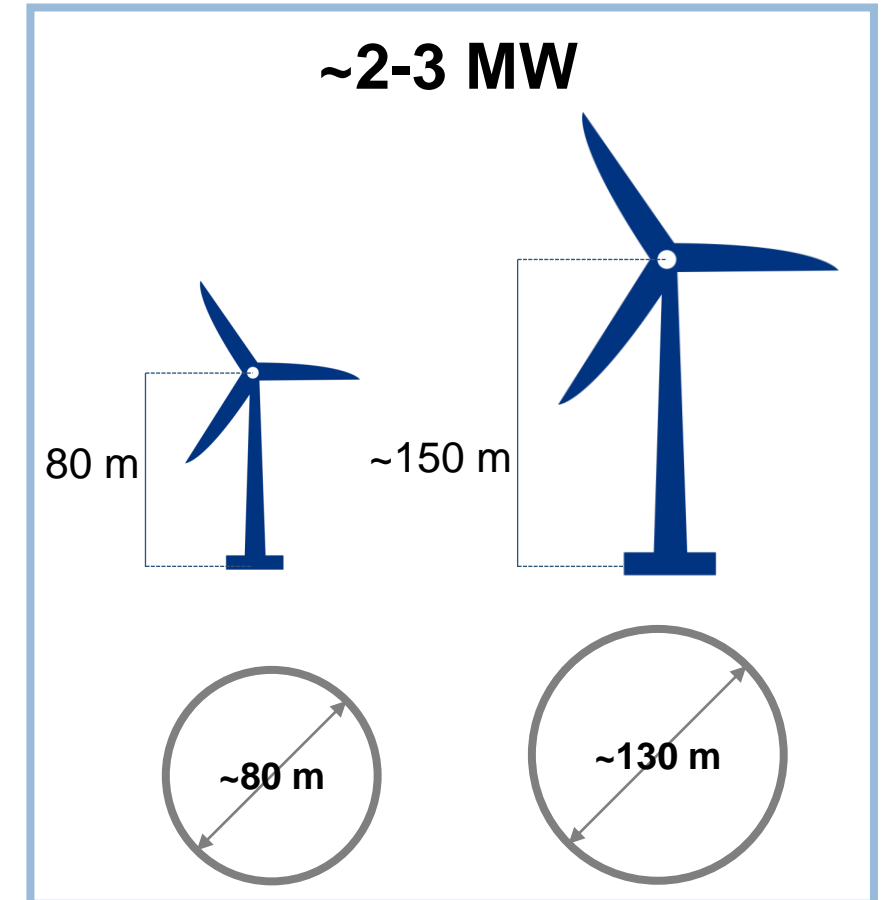


# The IceLoss 2.1 model chain



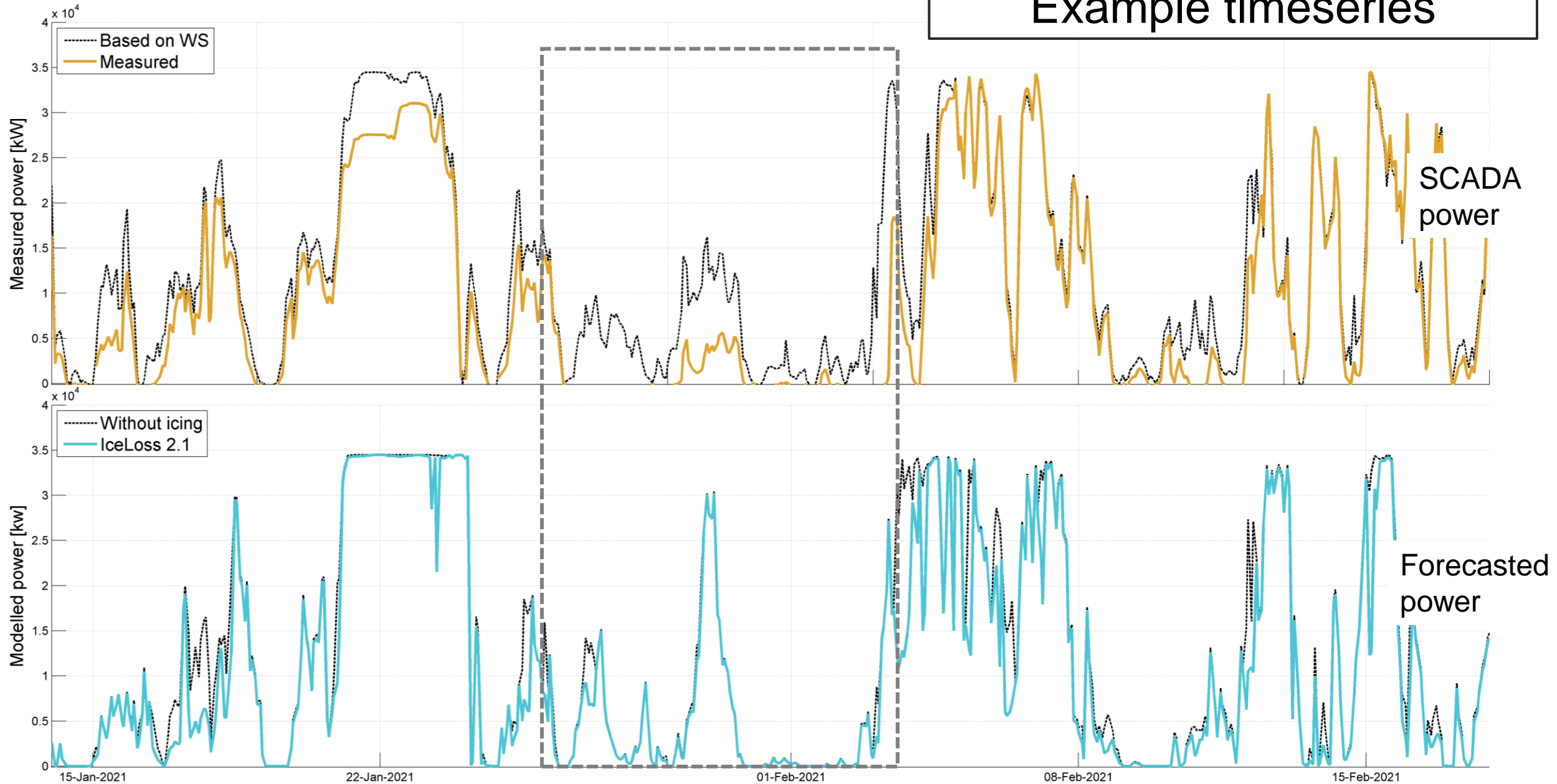
# IceLossForecast 1.0 running at four wind farms in Finland

- ▶ Four WFs close to the northern shores of the Gulf of Bothnia
- ▶ ~10 turbines per WF
- ▶ Ground elevation <120 m ASL, internal variation <12 m
- ▶ No ice protection system
- ▶ Forecasts ongoing since 11/2019
- ▶ The client especially interested in day-ahead forecasts for stops due to icing



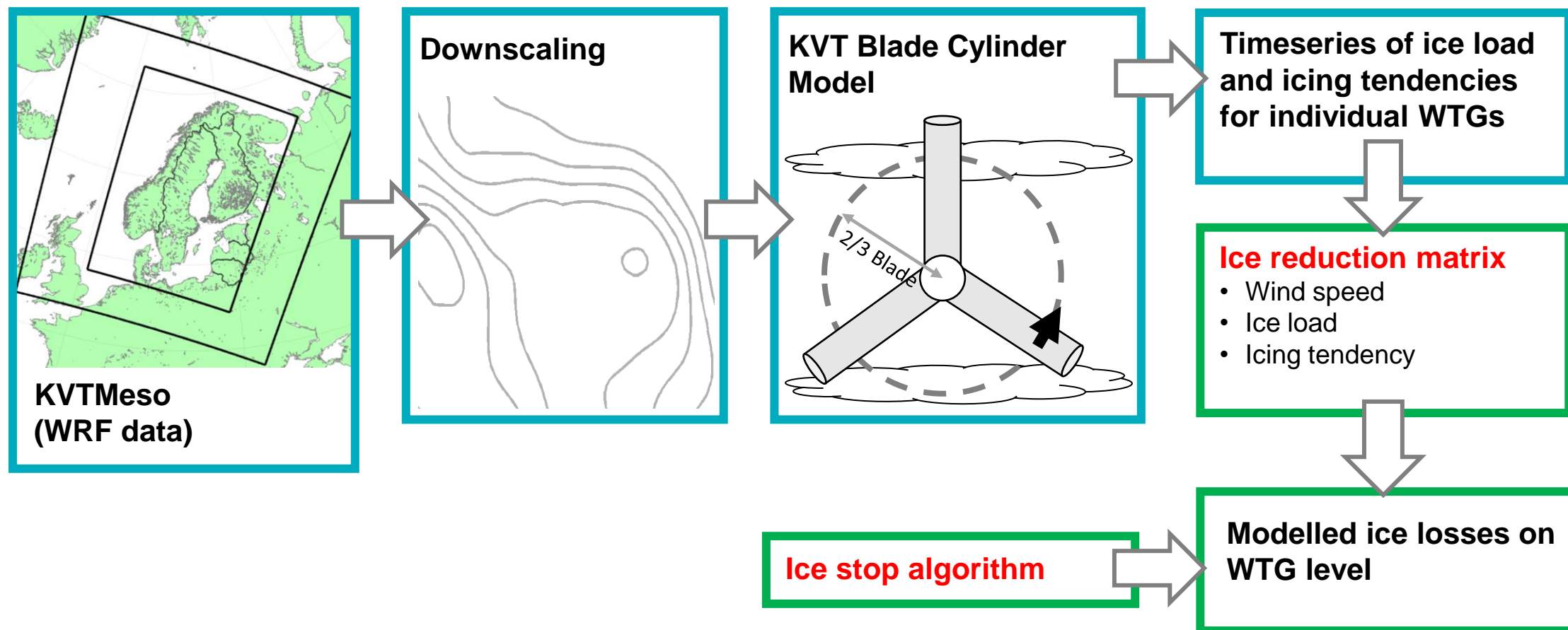


# Example timeseries

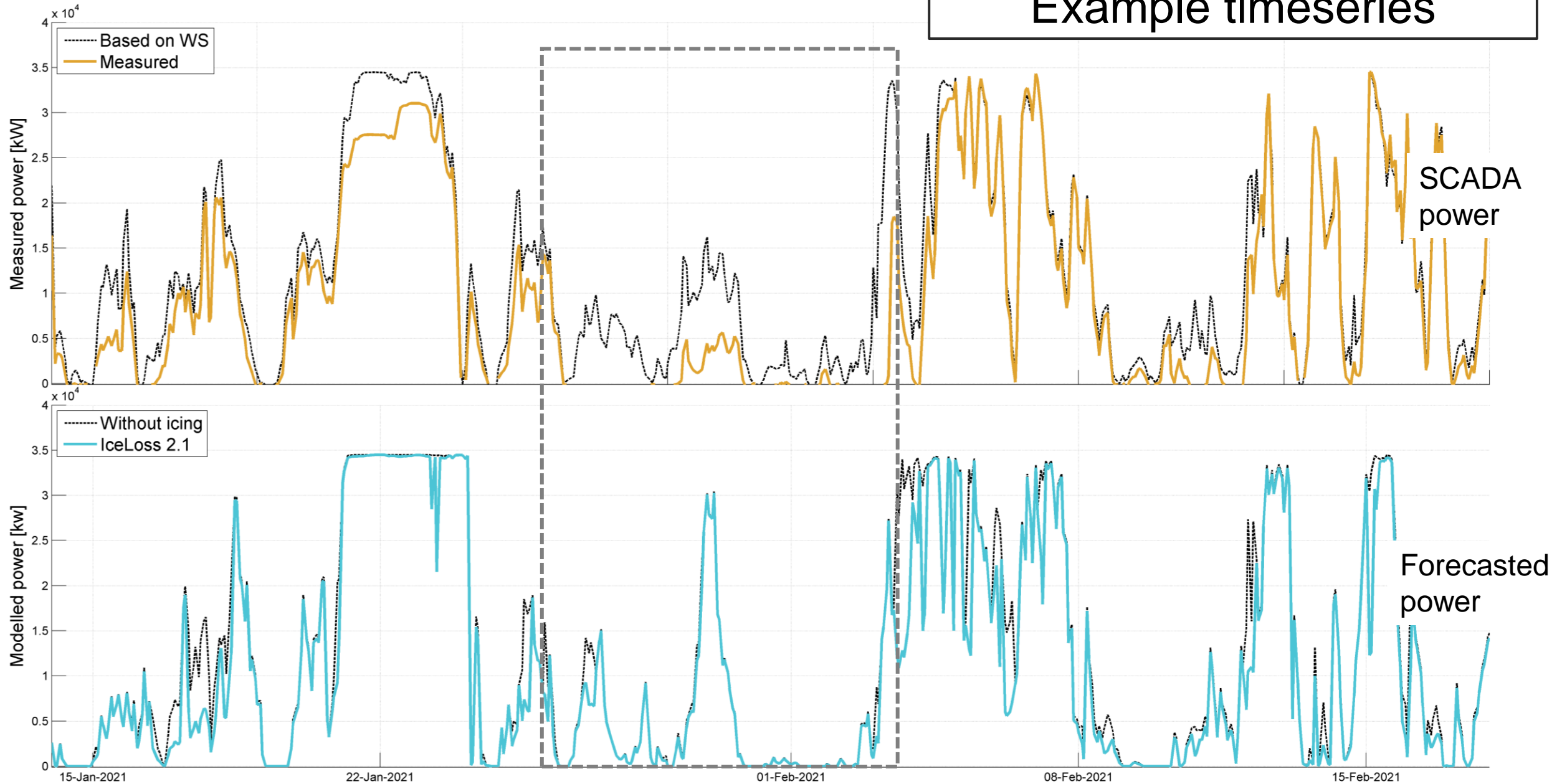




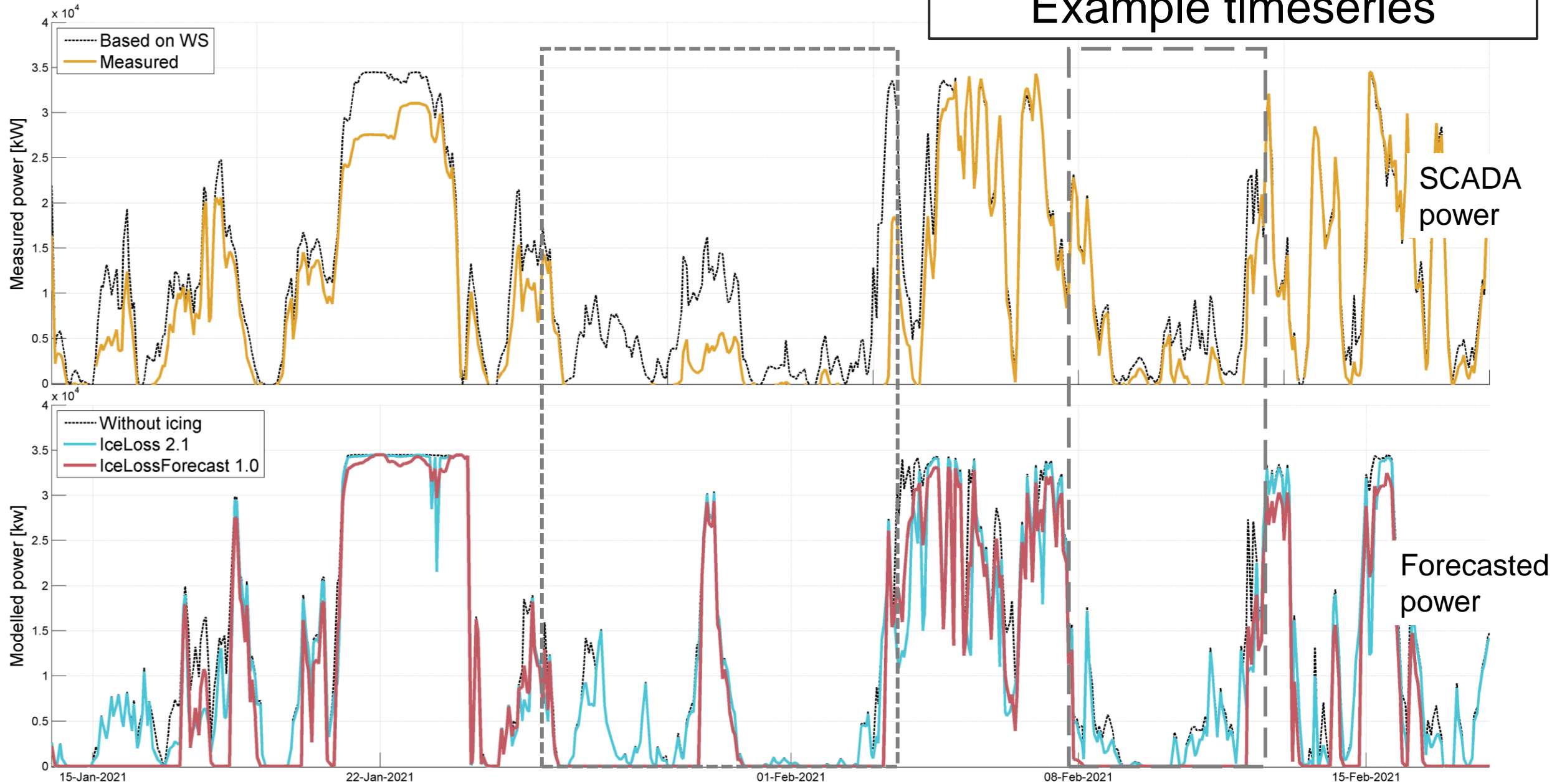
# The IceLoss 2.1 model chain



# Example timeseries



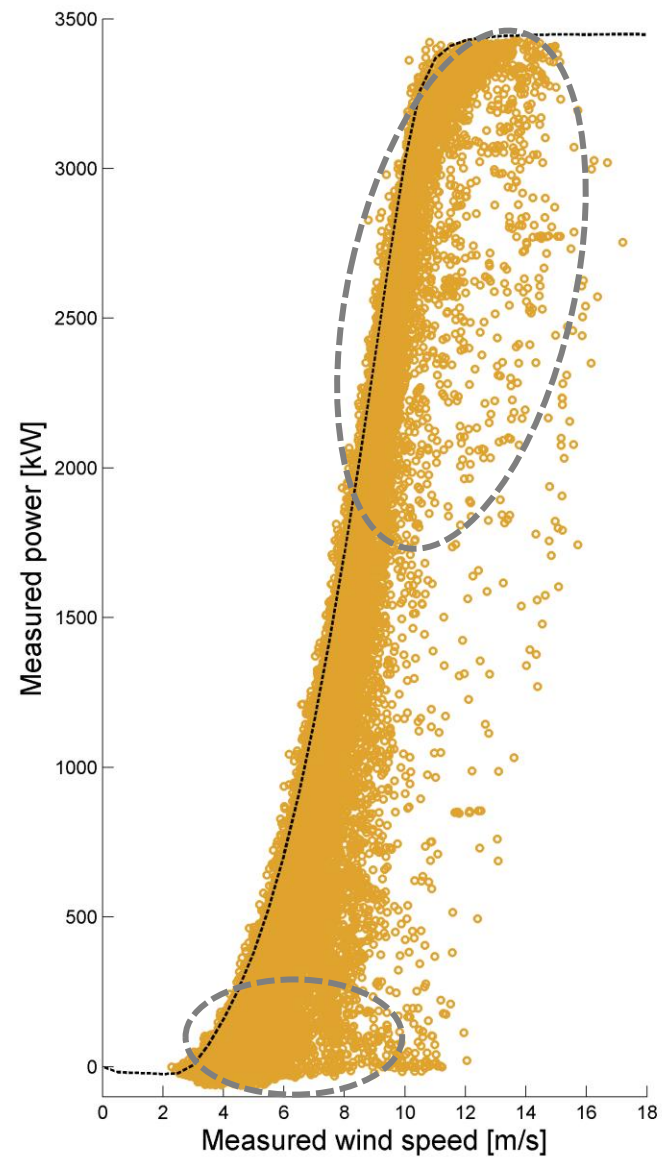
# Example timeseries



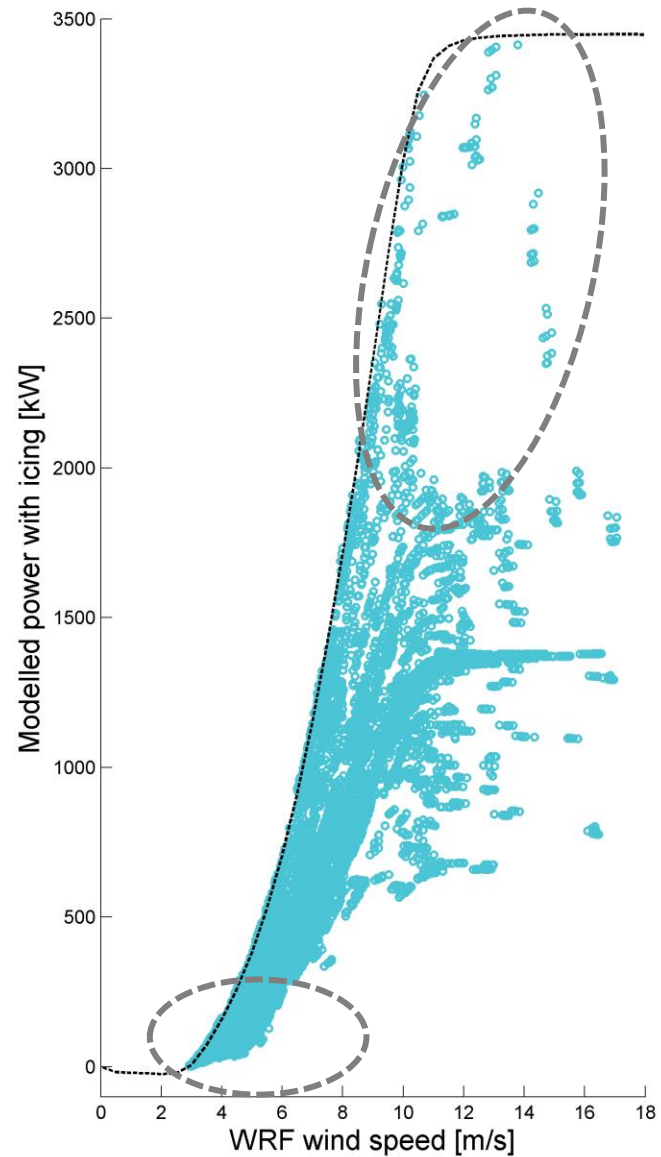
SCADA  
power

Forecasted  
power

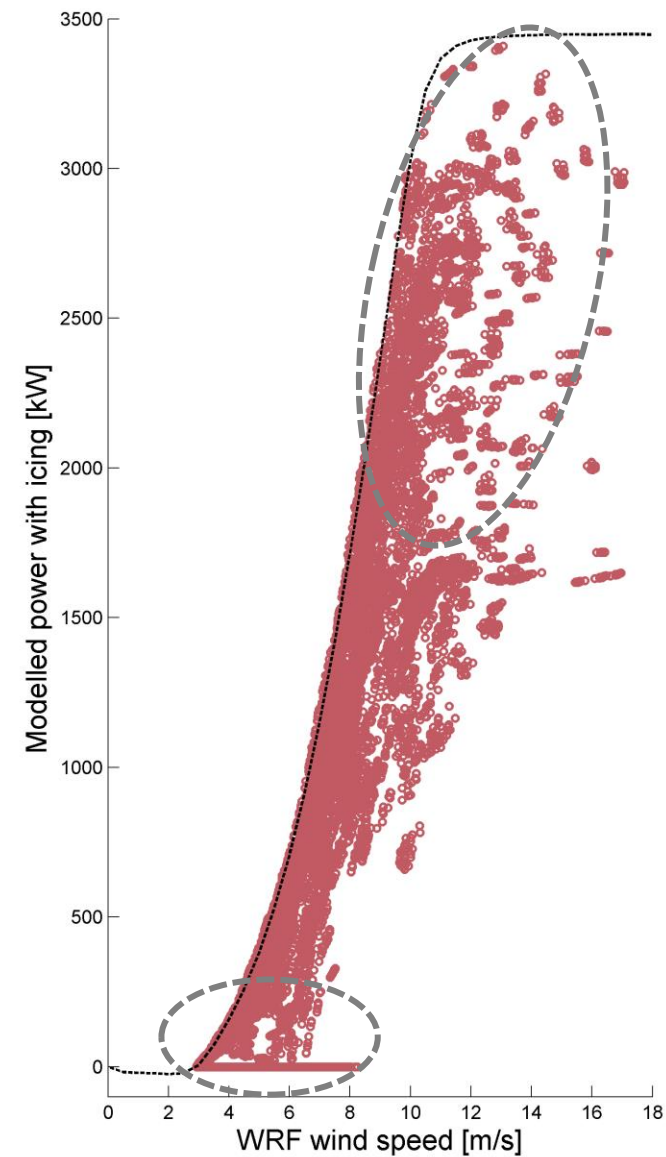
# SCADA



# Original



# IceLossForecast 1.0





# Hit rate for stops improved, but false alarms increase

	Stops due to icing		Total loss due icing	
Site	Hit rate	False alarm rate	Measured	Modelled
WF1	0% → 41.0%	0% → 4.1%	3.0%	3.6% → 4.2%
WF2	0% → 22.9%	0% → 2.2%	1.8%	1.9% → 2.0%
WF3	0% → 30.2%	0% → 2.7%	1.5%	1.0% → 1.8%
WF4	0% → 31.7%	0% → 3.1%	3.6%	2.6% → 3.4%

Improved  
Worsened

		Observation	
		Stop	No stop
Model	Stop	a	b
	No stop	c	d

- Hit rate:  $a/(a+c)$
- False alarm rate:  $b/(b+d)$

# Some observations and ideas for improvement

- ▶ Choose the correct metrics / cost functions in model optimization
  - ▶ MAE of total loss **vs.** hit rate for stops
- ▶ Parametrisations for long-term icing losses  $\neq$  parametrisations for short-term icing losses
  - ▶ Especially stops
- ▶ Incorporating SCADA data to the modelling to improve timing issues
- ▶ Machine learning based model:  
Option for a physical model?
  - ▶ Using WRF output as the model input



# Summary

- ▶ Day-ahead icing forecasts are crucial for electricity marketing pricing
- ▶ A prototype of an operational icing forecast model (IceLossForecast 1.0) is currently running at 4 wind farms in Finland
- ▶ Further development is ongoing
  - ▶ Tuning the IceLoss model parametrisations (ice reduction matrix)
  - ▶ Introducing a separate algorithm for stops
    - Found to improve the forecast already significantly
  - ▶ Incorporating real-time SCADA data in modelling → IceLossForecast 2.0

**Thank you!**



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