

IceWind Inter-comparison of Icing Production Loss Models

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IceWind Inter-comparison of Icing Production Loss Models

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 - Pierre Pinson (**DTU Electro**)
 - Mark Žagar (**Vestas**)
- Kjeller Vindteknikk
 - Øyvind Byrkjedal

- VTT
 - Timo Karlsson
 - Tomas Wallenius
 - Ville Turkia
- WeatherTech Scandinavia
 - Stefan Söderberg
 - Magnus Baltscheffsky



IceWind

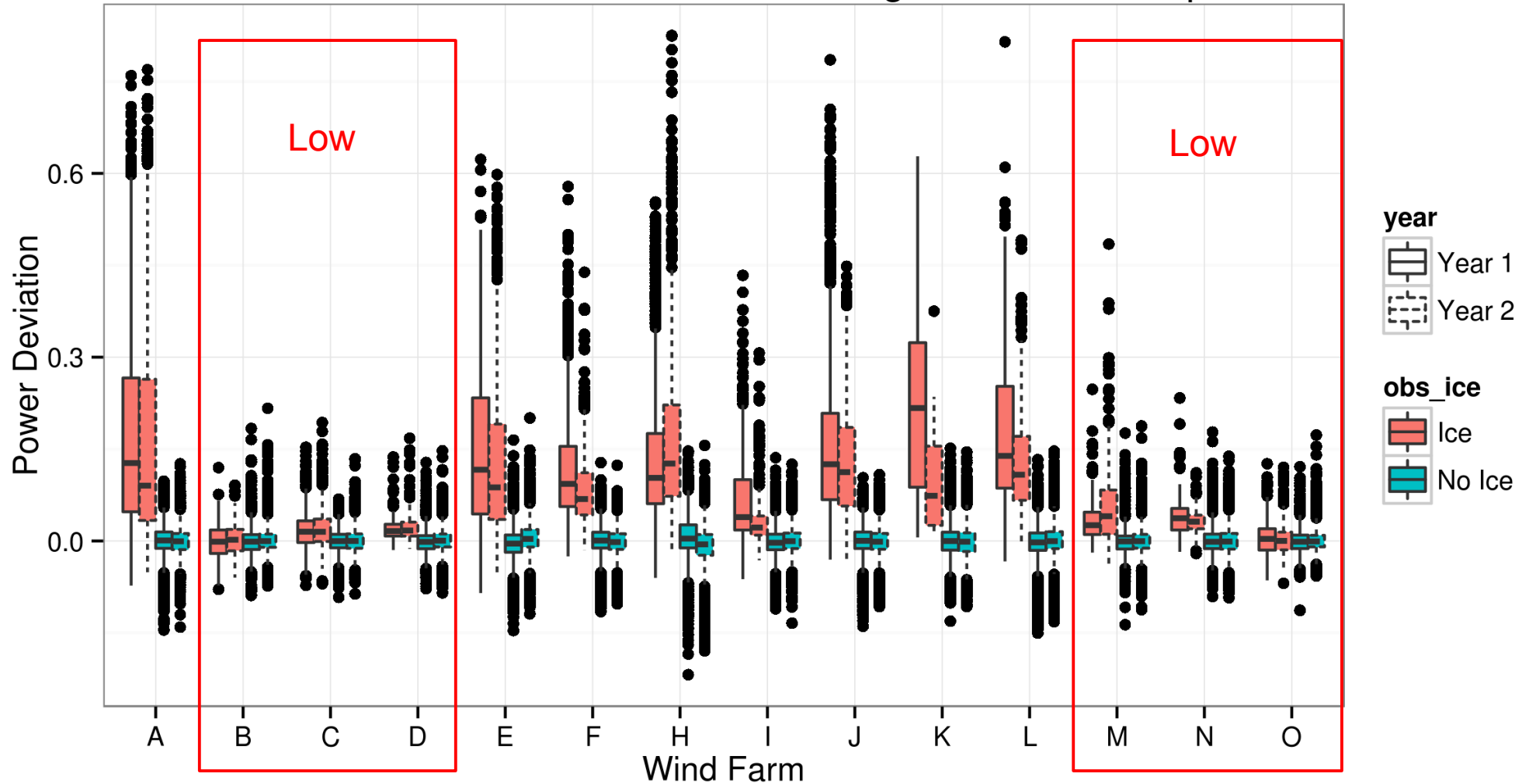
- Nordic Project supported by Top-Level Research Initiative (TFI)
- Improved forecast of wind, waves and icing
- 13 Project partners
- Work Package 1: Wind turbine icing

Observations

- Selected data from 15 wind farms
 - Averaged to wind farm values, not turbine specific
 - 2 years of data (June 2010-June 2012)
 - Observed icing times from automated approach classifying production loss
 - Data removed when turbines not operating optimally



Power Deviation from Power Curve Using Nacelle Wind Speed



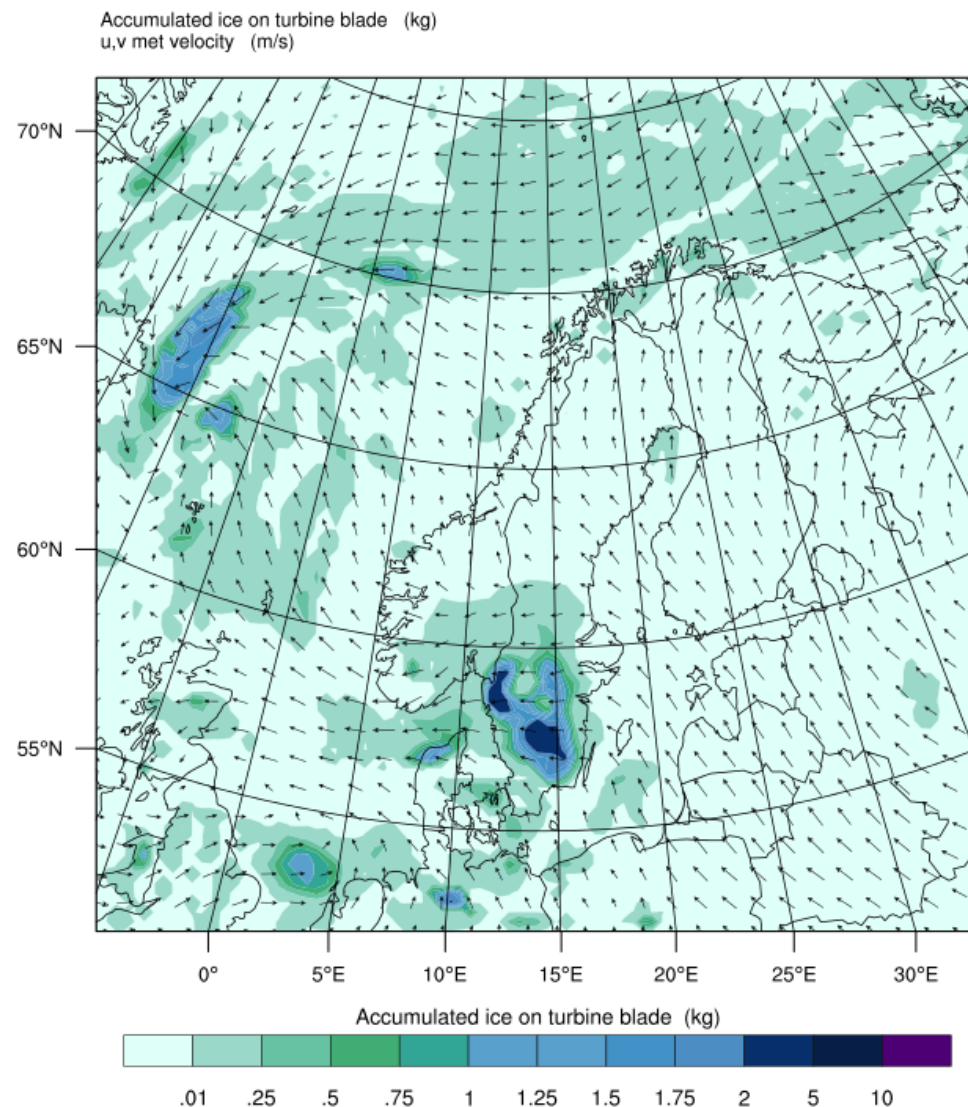
- Note 2 regimes in different wind farms
- Similar results from both years

WRF model data

Turbine Icing @ 80m

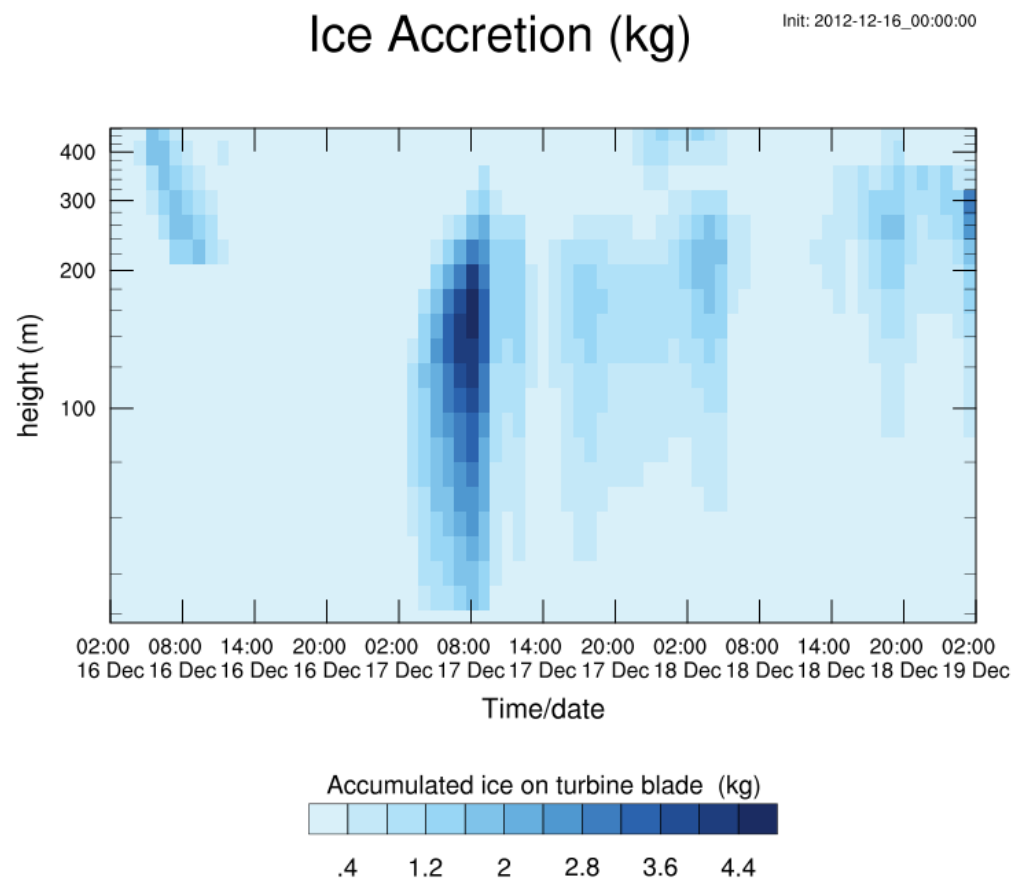
Init: 2012-12-16_00:00:00
Valid: 2012-12-17_17:00:00

- Provided by Vestas at 3 km
- WSM5 microphysics
- 6 hour spin up cycle
- Provided Fields
 - Wind Speed
 - Temperature
 - Pressure
 - 4 Cloud types
 - Precipitation rate
 - Specific humidity
 - Shortwave radiation
 - Longwave radiation



WRF model data

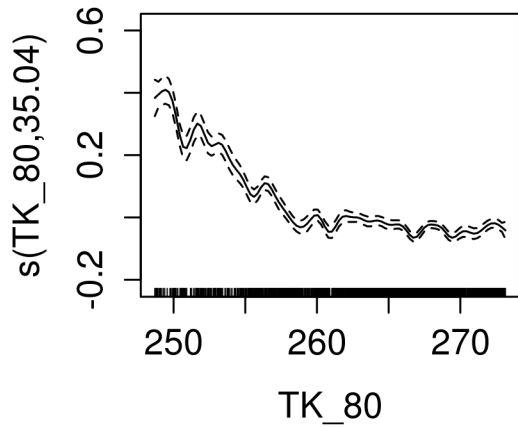
- 3D Representation of the atmosphere
- Data interpolated to 40, 80, 120, 160 and 200 m AGL



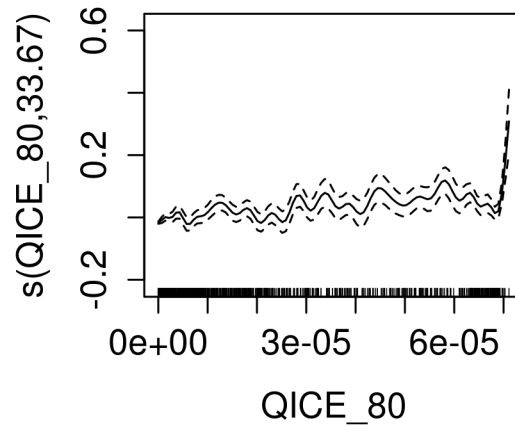
Production Loss Models

DTU model

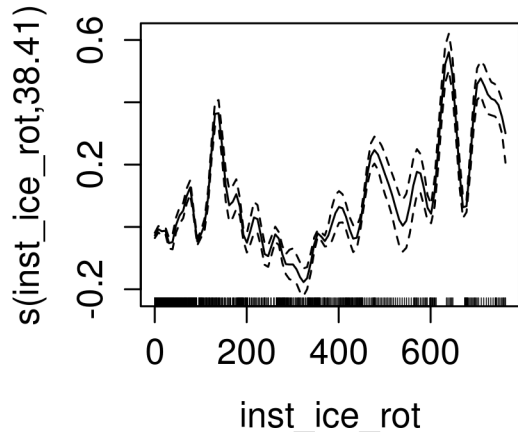
Farm A



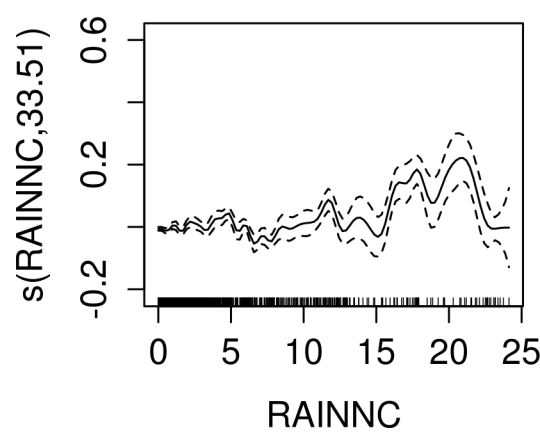
Farm A



Farm A



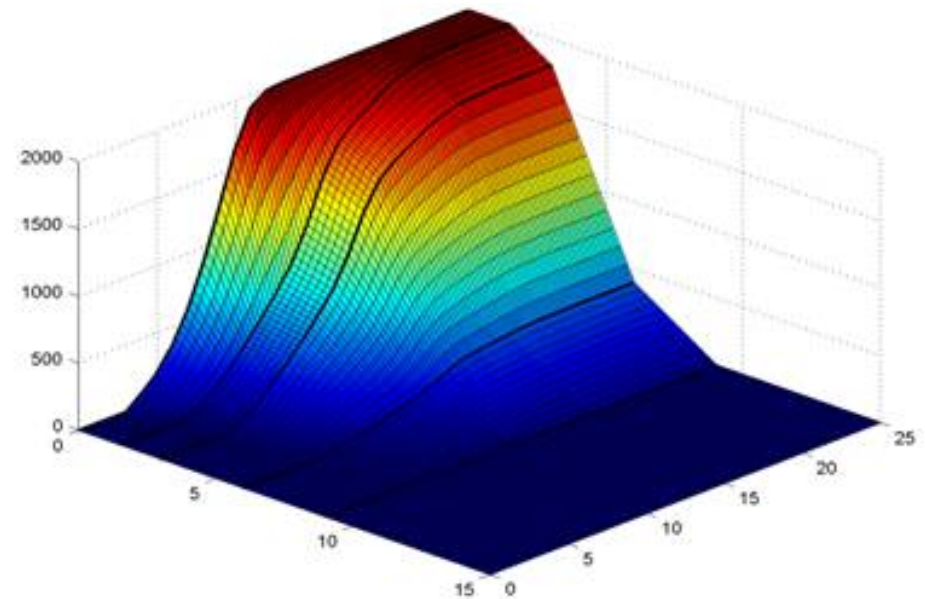
Farm A



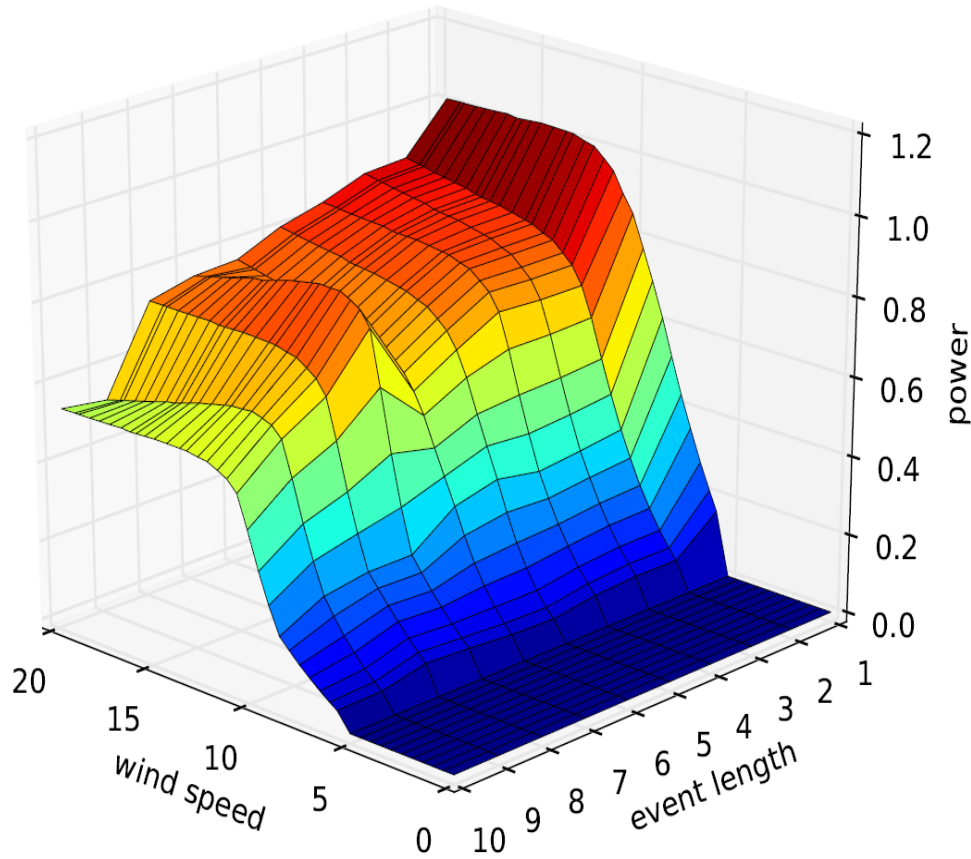
- Mixed model
 - Fits separate models for forecast ice / no ice conditions
- Generalized Additive Model
- Utilizes results from WRF and iceBlade
 - IceBlade modified to include cloud ice for WSM5 microphysics
- Fit separately for each farm in this study with consistent variables

Kjeller Model

- Two-parameter power curve
 - Suggested by wind tunnel results
 - Ice mass and wind speed
 - Tuned and validated using operational data
- Uses a standard cylinder for ice mass modeling
- Assumes power yield of 0 at approximately 9 kg/m



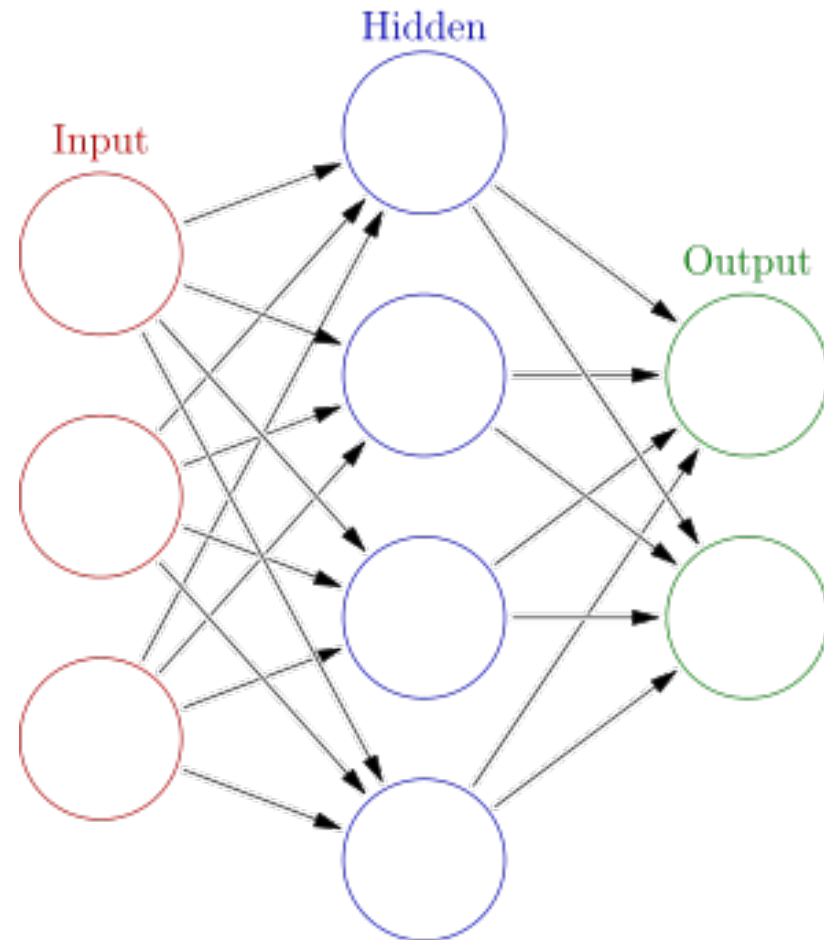
VTT



- Based on statistical analysis of power loss observations
- Produces an estimate for power loss due to rotor icing
 - Based on wind speed and length of icing event
- Independent of icing or production forecasting methods
- Requires external icing forecast
 - Used iceBlade accumulated icing for this comparison

WeatherTech Scandinavia

- WICE – WeatherTech Ice Model
- Artificial Neural Network
 - Trained with observed clean & iced production
- Tested for different turbines & locations
- Either forecast or assessment tool

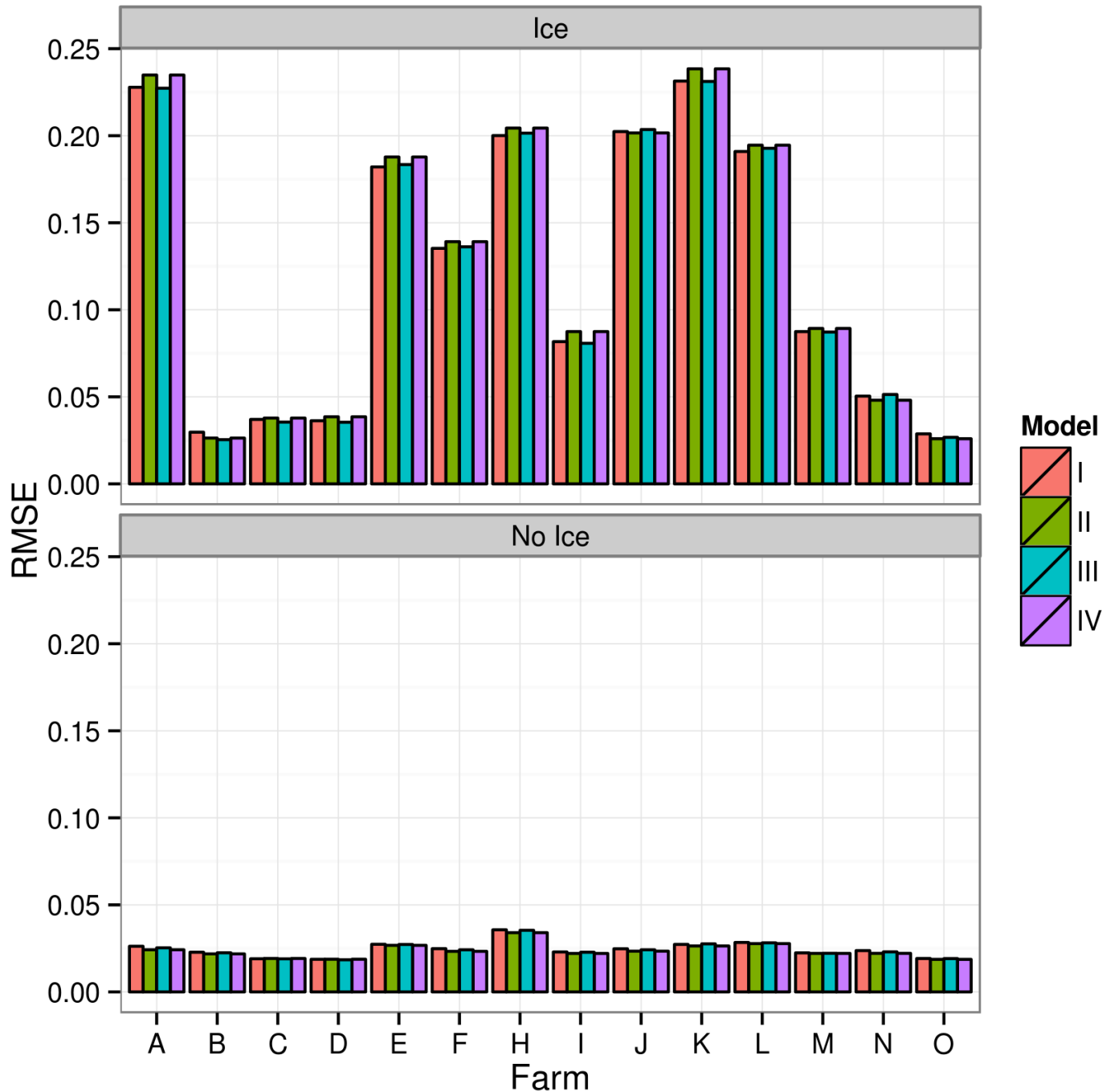


Results

Terminology

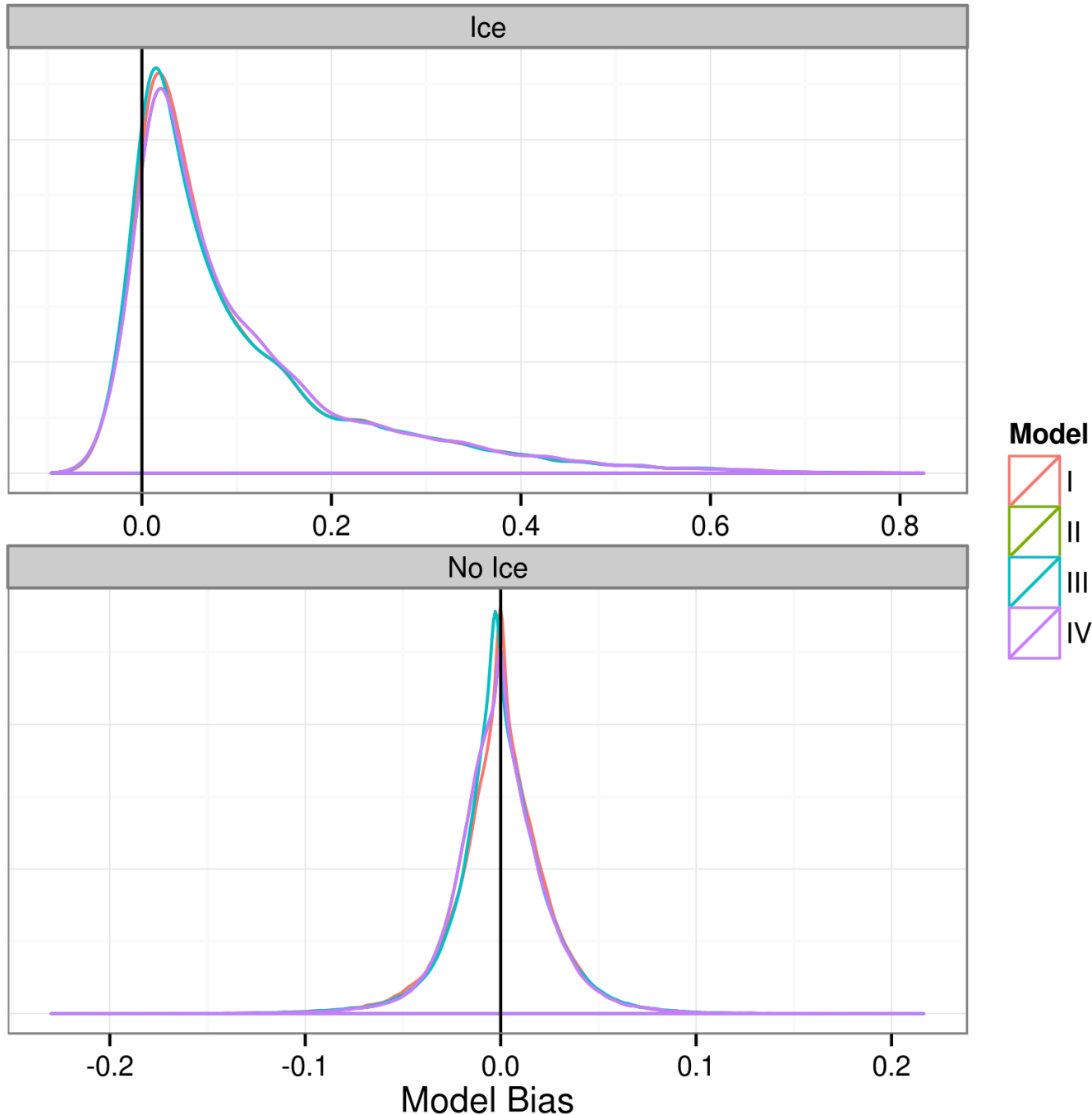
- 2 Years (defined June – May)
 - Year 1 Used to fit statistical models
 - Year 2 Evaluation year
- 2 Power estimates
 - Gross: power estimate without icing
 - Iced: power estimate with icing
- 2 Observed Conditions
 - Ice: times when observations suggest icing
 - No Ice: times without icing
- 14 Farms (Labeled A-O, ex. G)
- 4 Models (Labeled I-IV)

Gross Power RMSE



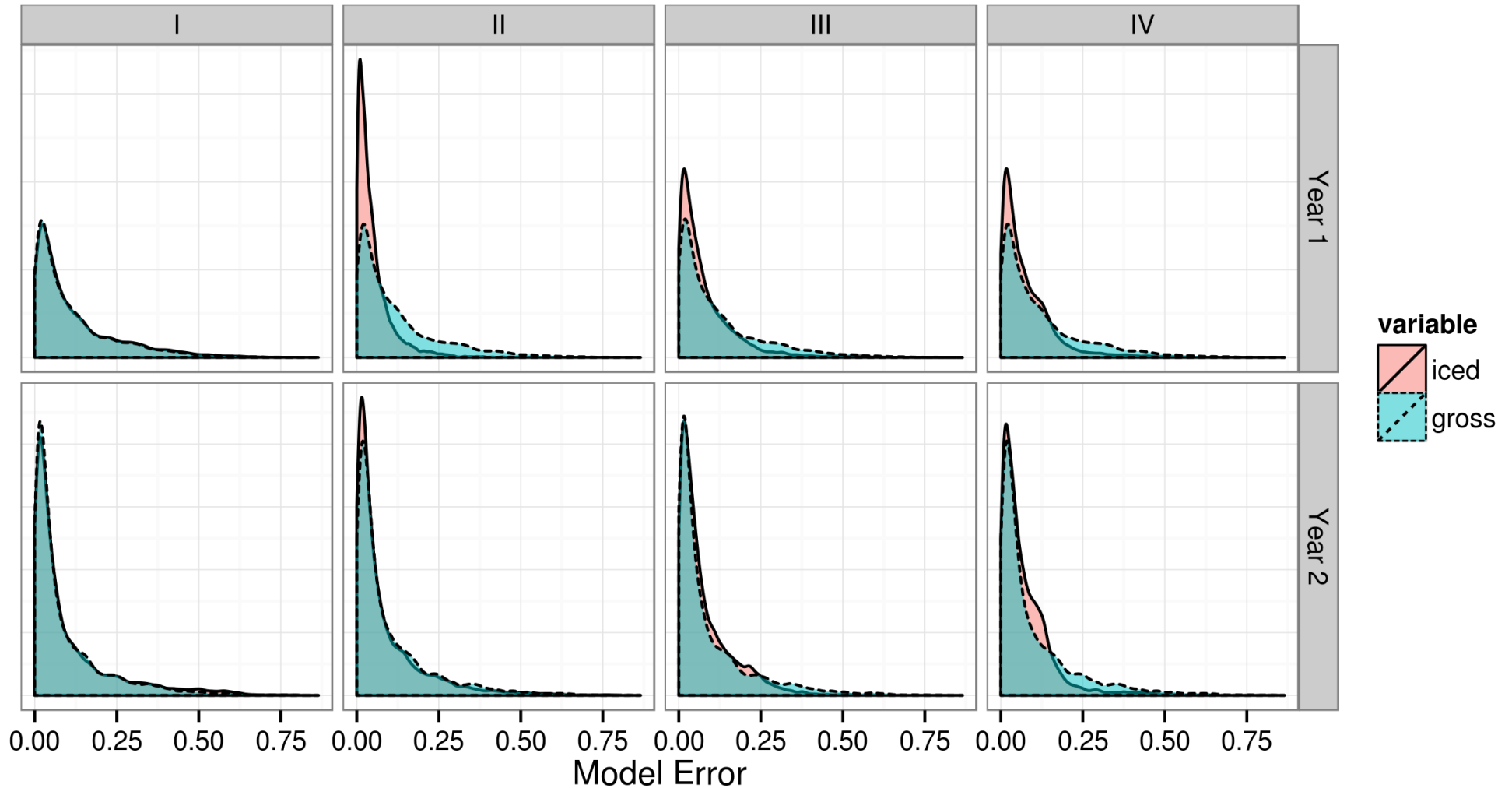
- Power curve fit to nacelle wind speed
- Gross estimates similar across models
- Much larger errors for observed icing cases
- Error pattern similar to impact of icing from boxplot

Distribution of Gross Power Bias



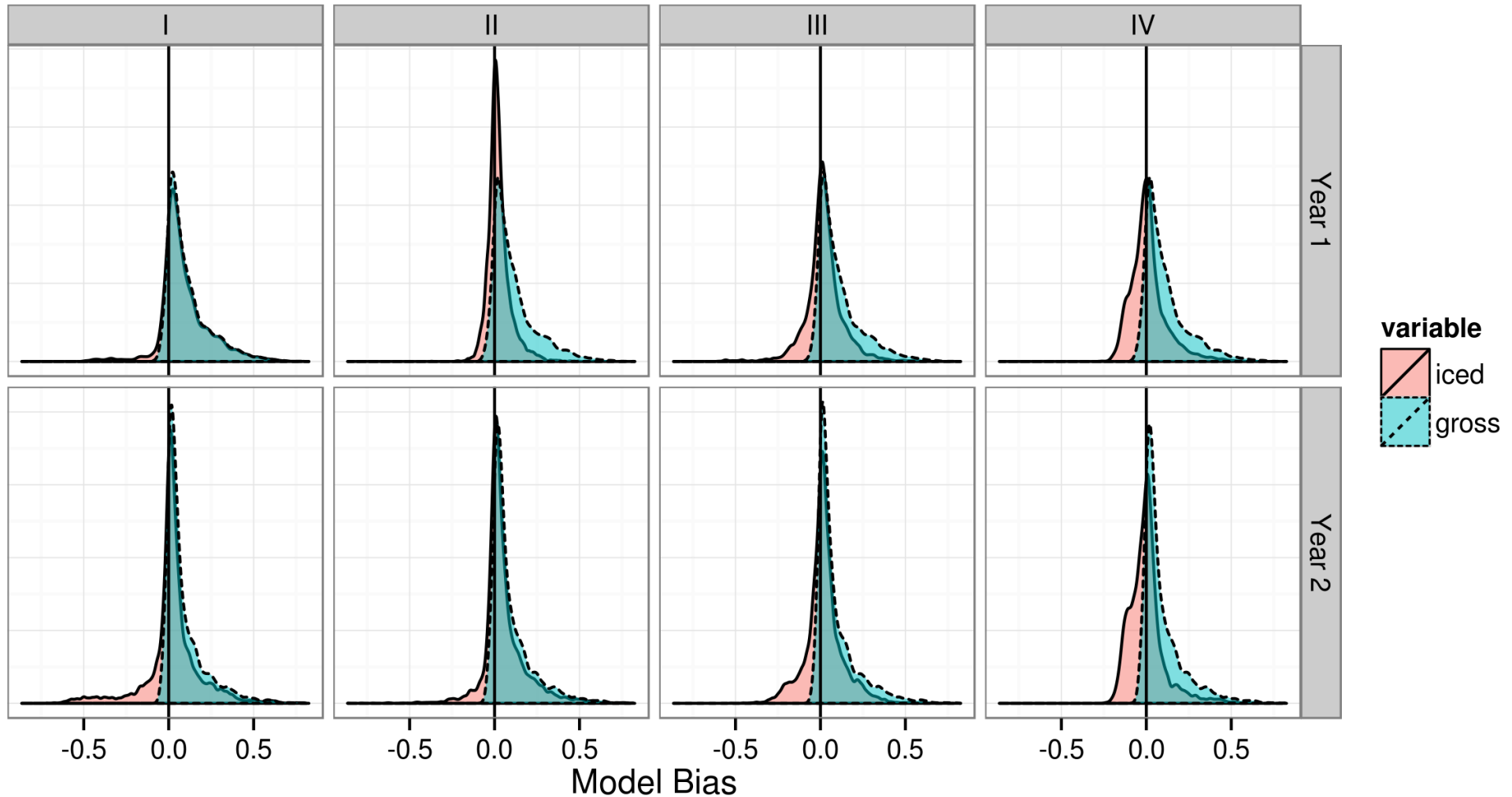
- Peak near zero for all models
- Symmetrical bias for no ice
- Ice condition skewed positive signifying higher estimated power than observed
- No large deviation across models

Iced vs Gross Error for Iced Times



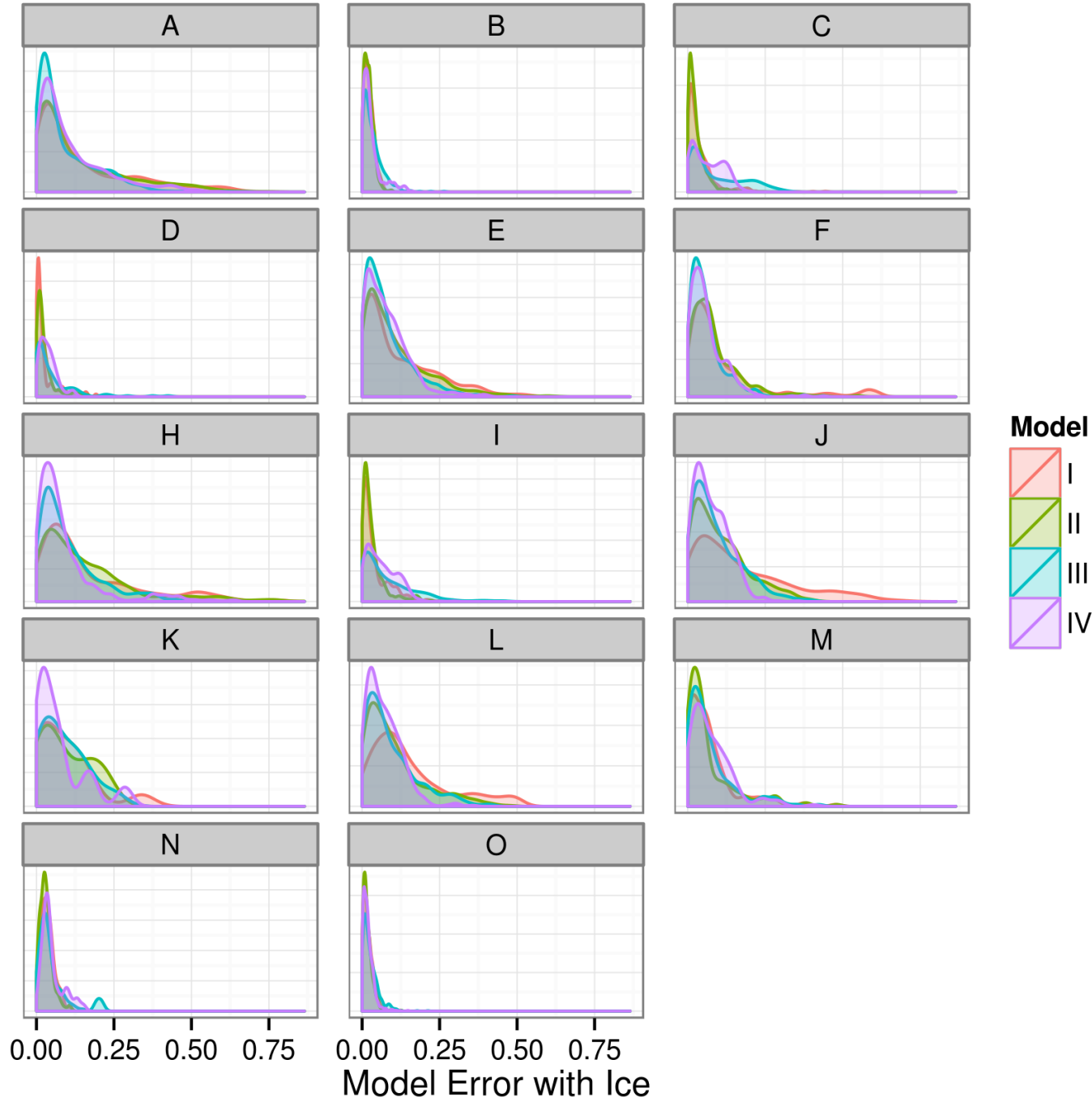
- Large improvement in year 1 for Models II, III and IV
- Much smaller improvement in year 2

Iced vs Gross Bias for Iced Times



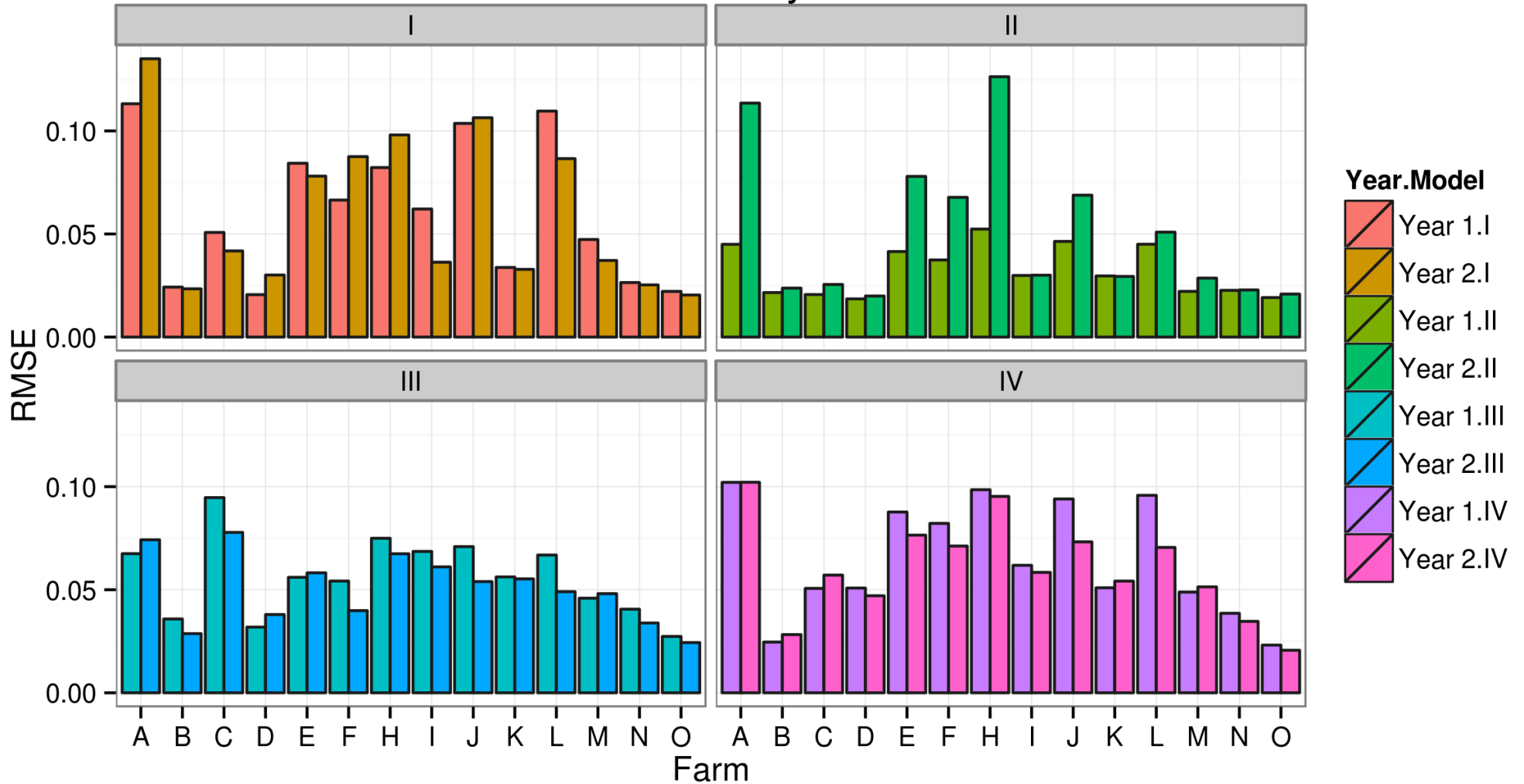
- General shift of positive bias to negative
- Year 2 shows larger shift of bias from positive to negative

Modeled Error for Year 2 Iced Times at Each Wind Farm



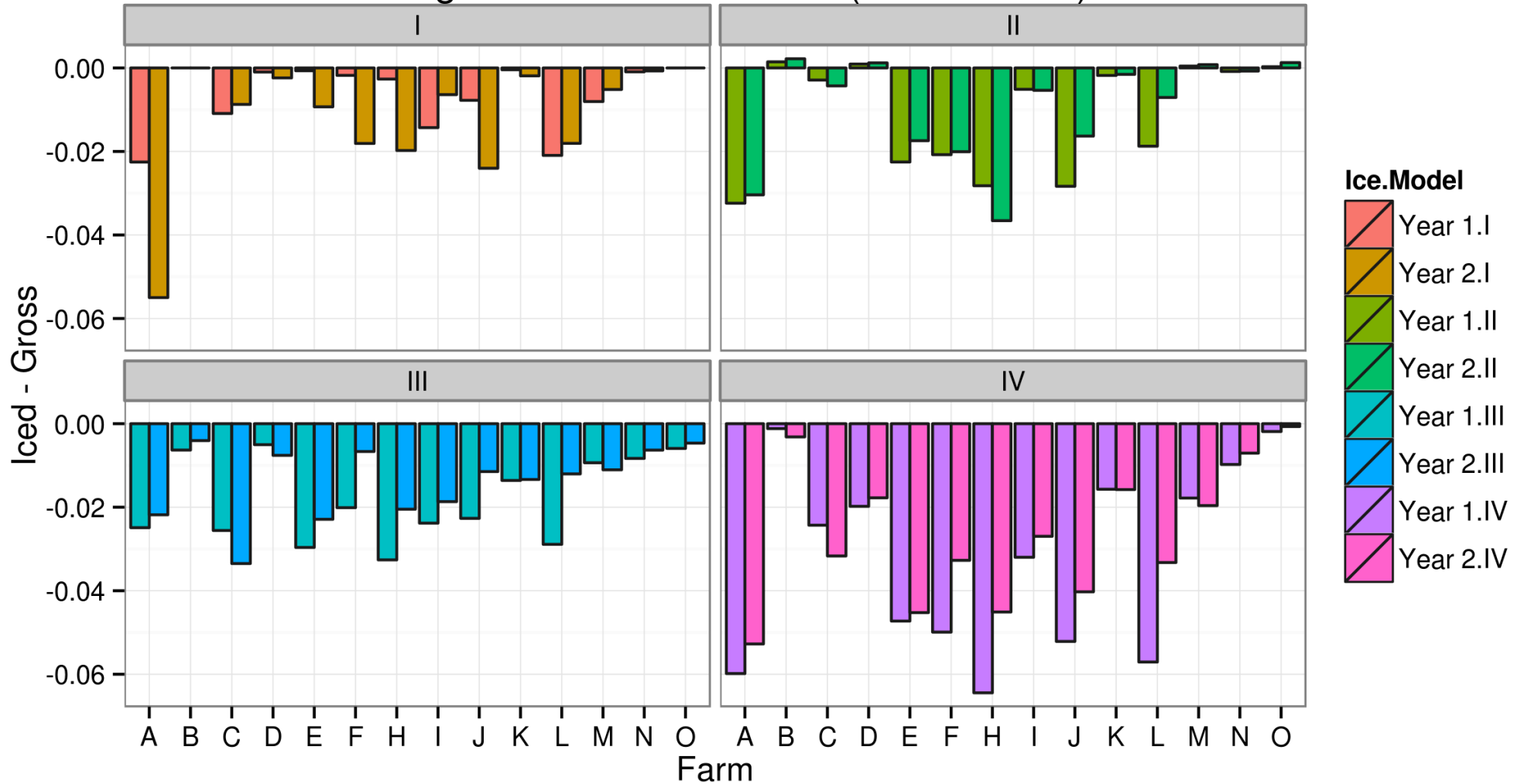
- Reasonable performance from all models
- Large differences between models at most sites
- Can pick out sites with low ice impact
- Model III and IV slightly outperform other models at several sites

Model RMSE by Year



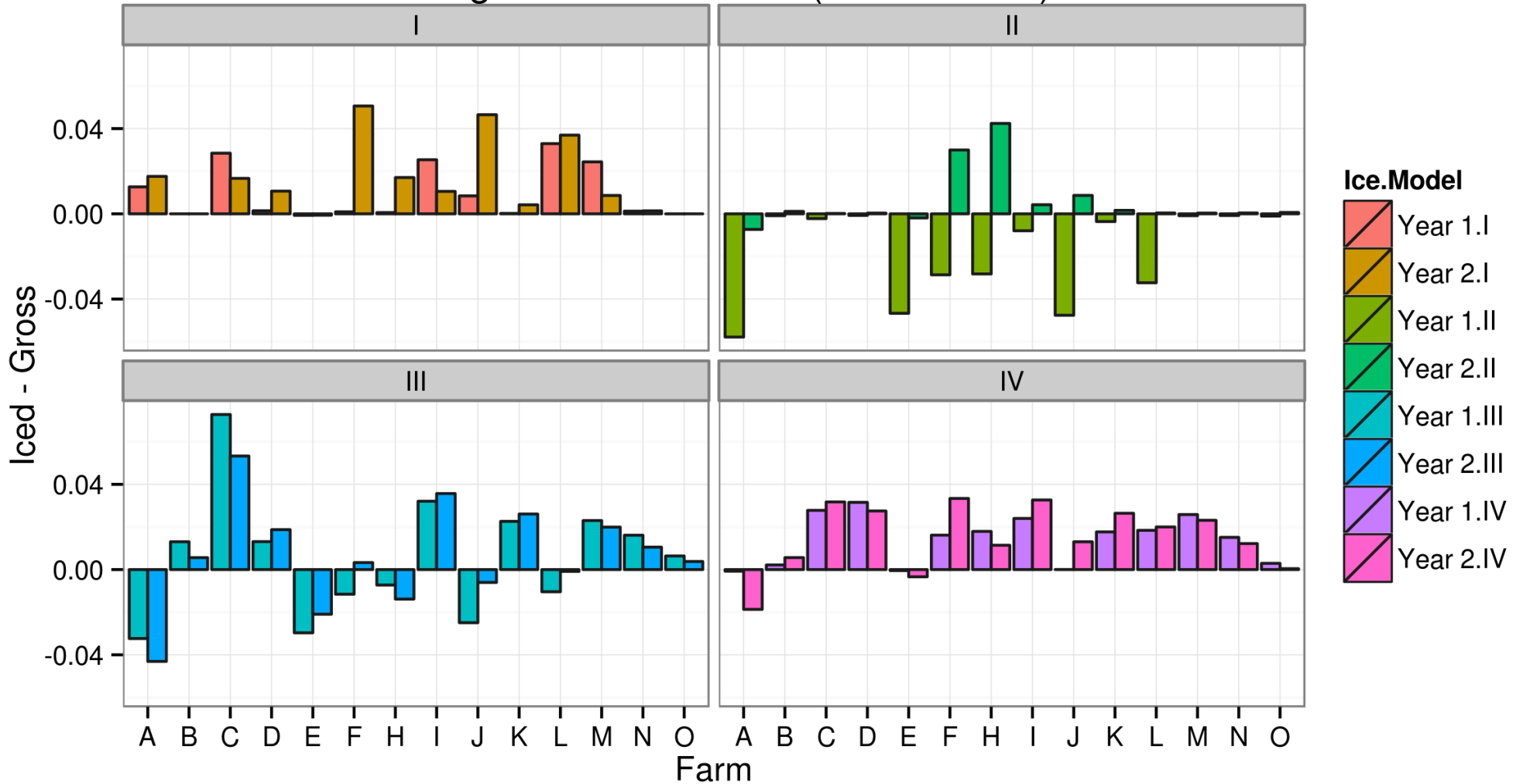
- Colors signify years
- Model II appears to have over fit model to year 1
- Models III and IV shows larger errors at sites with less icing than other models

Change in Model Mean Bias (Iced - Gross)



- Using any ice model almost always reduces bias
- Bias reduced more for sites with large amounts of ice
- Not a large change from year to year

Change in Model RMSE (Iced - Gross)



- Iced RMSE often worse than Gross, due to the decrease in performance for non-iced times
- Depending on agreement, bias correction may offset increased error

Conclusions

- Models perform similarly
- Differences appear mostly due to park conditions
- Large improvements still possible
 - Longer periods for model fit to reduce over fitting
 - WRF runs customized for icing
 - Ice ablation methods & relationship to power
- Agreed upon metric is needed to help improve the models
 - Bias was improved at most sites
 - RMSE was not improved as much
- Using human input could improve these models, need judgment on when to apply them