



## **Wake Losses in Wind Plants – Comparing Several Methods with Measured Wakes**

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## Why?

- **Wind farm performance typically is less than what was projected during the wind resource assessment, and turbine (SCADA) data cannot fully explain underperformance.**
- **Traditional wake estimation conceptually based on single wake; large wind farms have more of an aggregate effect.**
- **Traditional wake models assume wake recovery is diffusive process only.**

## Data Sources

- **Meteorological towers (before and after construction)**
- **Wind resource assessments (gridded data)**
- **Gridded reanalysis data (MERRA)**
- **Wind Turbine (SCADA) data – wind speed, power, curtailments, turbine faults, etc.**

## Models

- **Traditional wake models (WindFarmer, OpenWind)**
- **WindFarmer, OpenWind with large wind farm effects**
- **Weather Research and Forecast (WRF) model with turbine parameterization (Fitch et al. 2012)**

## Three Sources of Wake “Truth”

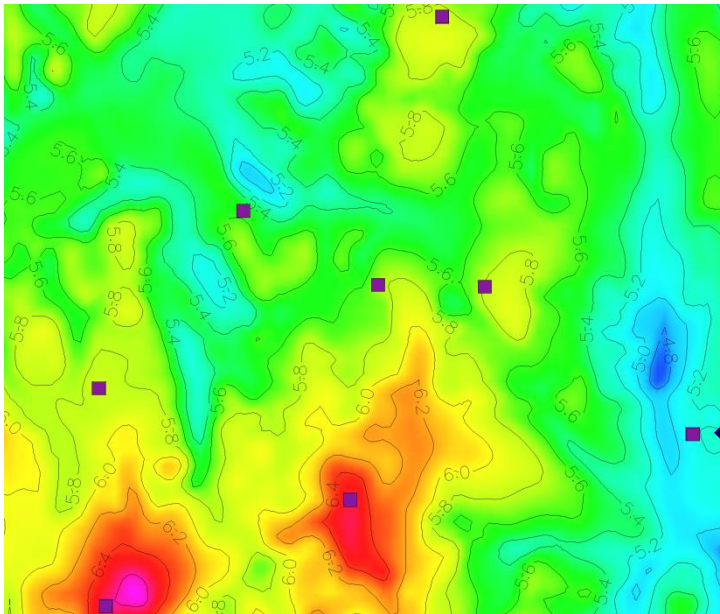
- **Upwind/downwind turbine pairs (SCADA data)**
- **Pre-post build tower analysis (Met towers and MERRA)**
- **WRA minus OpA (Modeling/Towers and SCADA)**

## WRA minus OpA

- Inherently a full-farm estimation
- By definition, aggregated over the fleet, the right amount of “wake” (WRA-OpA wake) loss that is needed to reflect the losses incurred
- WRA-OpA wake includes actual wake, flow deviation, and power curve inefficiencies due to effects such as shear, as well as inaccuracies in WRA speed estimates
- The above factors contribute to scatter in the WRA-OpA estimates for individual sites

## Wind Resource Assessments

- Conduct one year of WRF modeling
- Mass-constrained interpolation (Swift)
- Correct biases with meteorological tower data
- Long term normalization with MERRA dataset

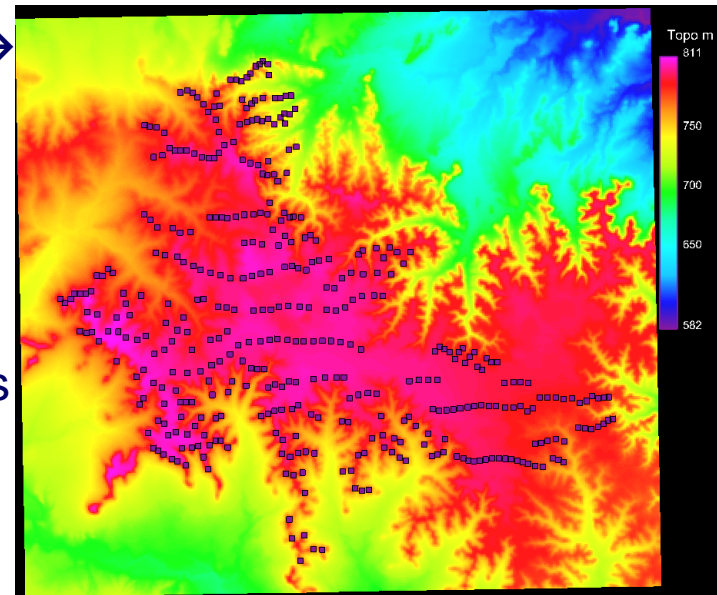
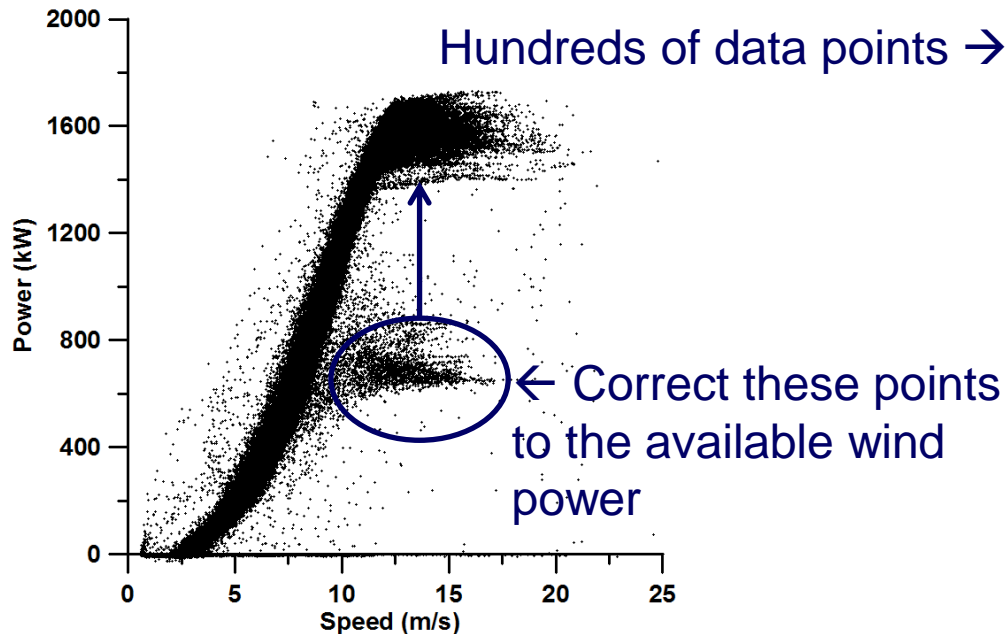


Map of wind resource without turbine wakeing is used as a baseline against which we compare operational data.

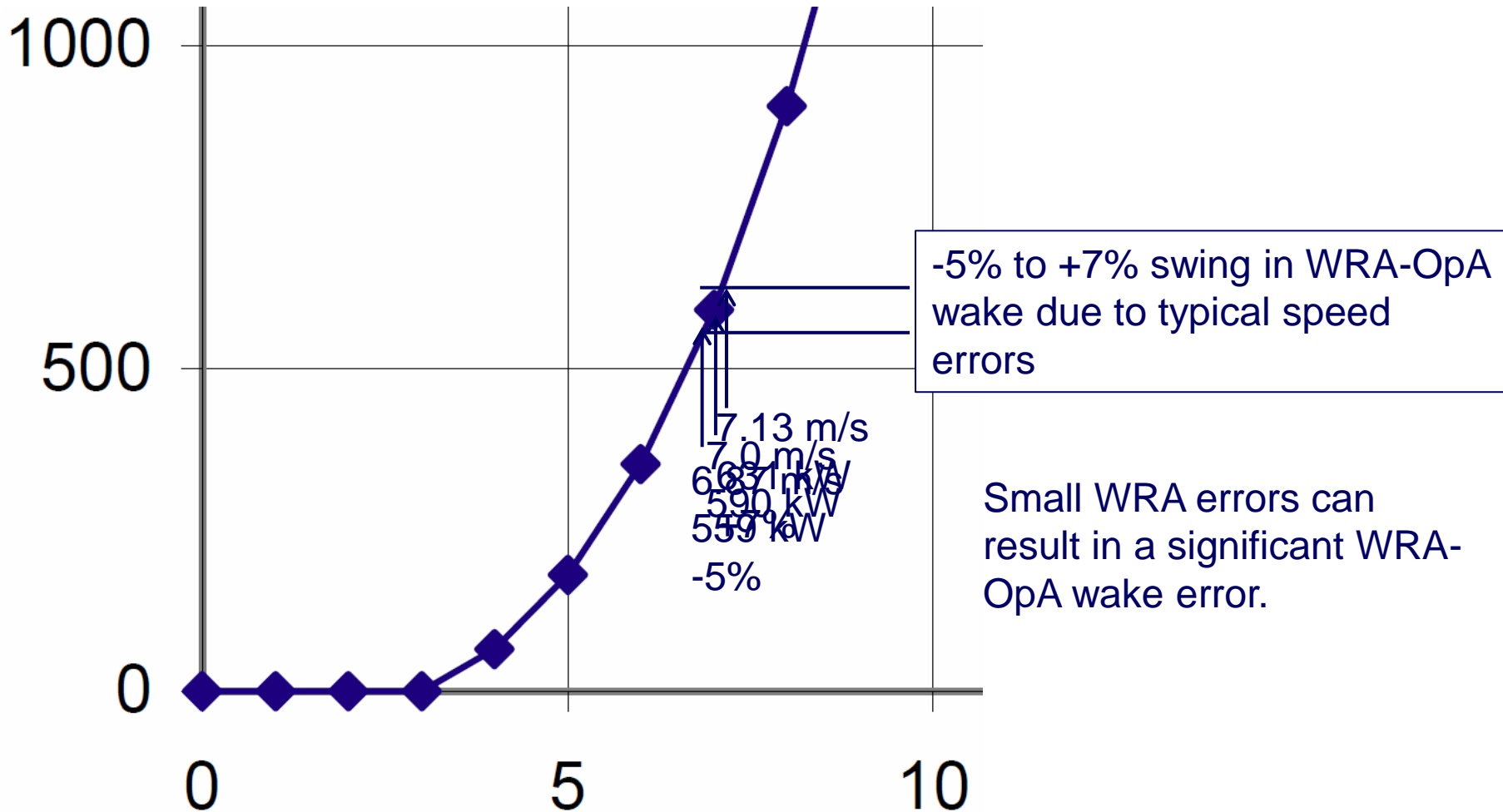
← Tower locations

## Operational Assessments

- Use SCADA data to measure wind speed and power.
- Operational power curve specific to each turbine.
- Correct data points where power is below normal curve—accounts for curtailments, operating faults, icing, etc.
- Result is available power with curtailment and availability losses removed, so remaining loss is due to wake.
- Normalize the result with long term (MERRA) data.

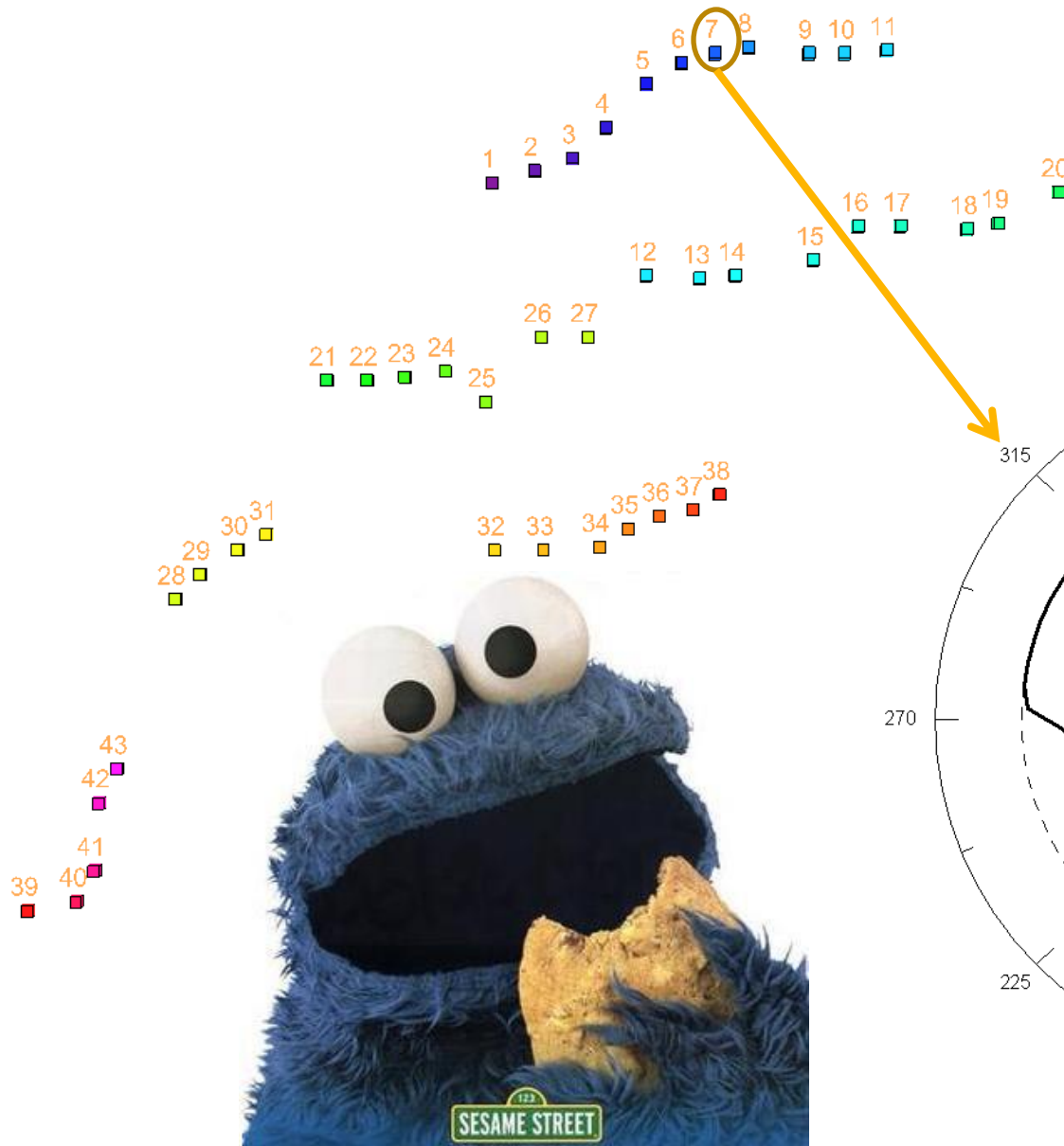


## WRA GCF Speed Sensitivity





# Upwind/Downwind Pairings

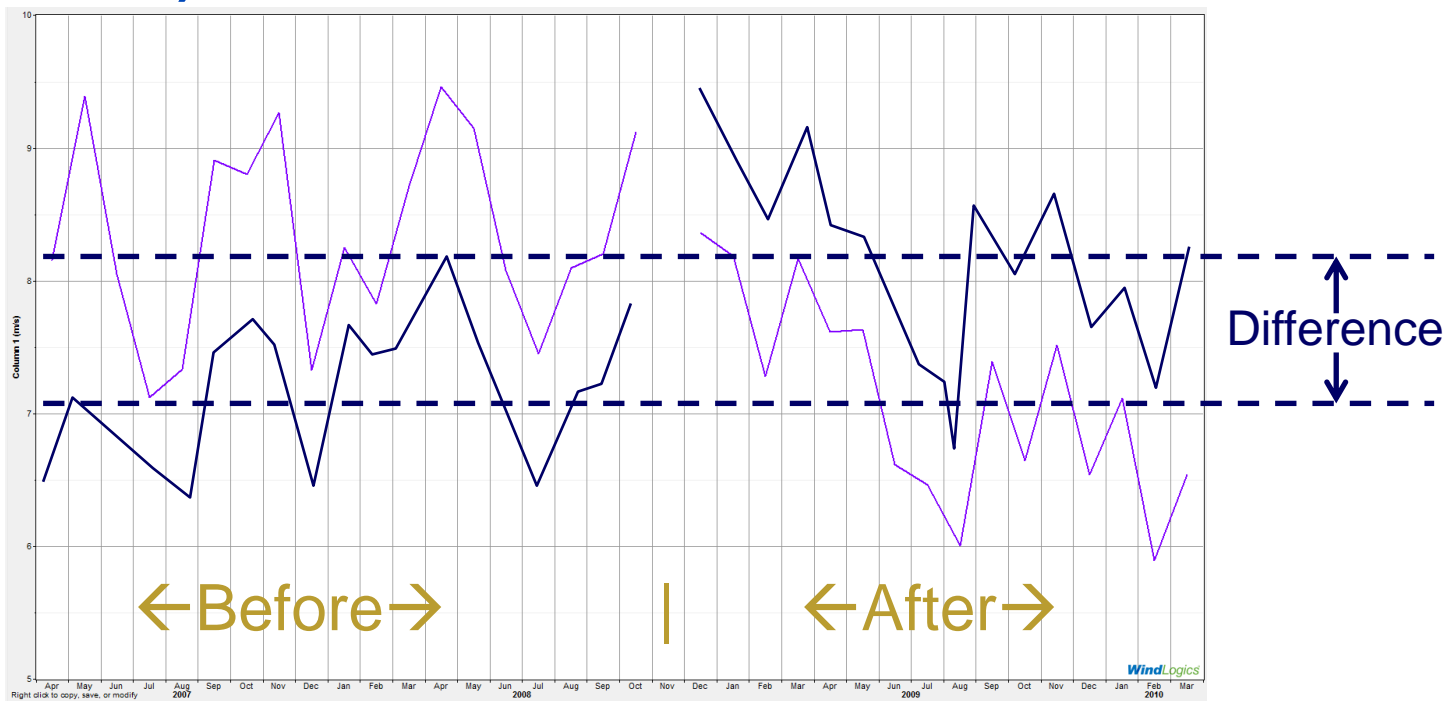


## Upwind-Downwind Turbine Pairing

- Works well for near-field pairings ( $< 10$  R. D.)
- Cannot see far-field wakes ( $\geq 10$  R. D.), thus sees only 2% or 3% loss per site
- Essentially abandoned as a method (only 3 sites analyzed)

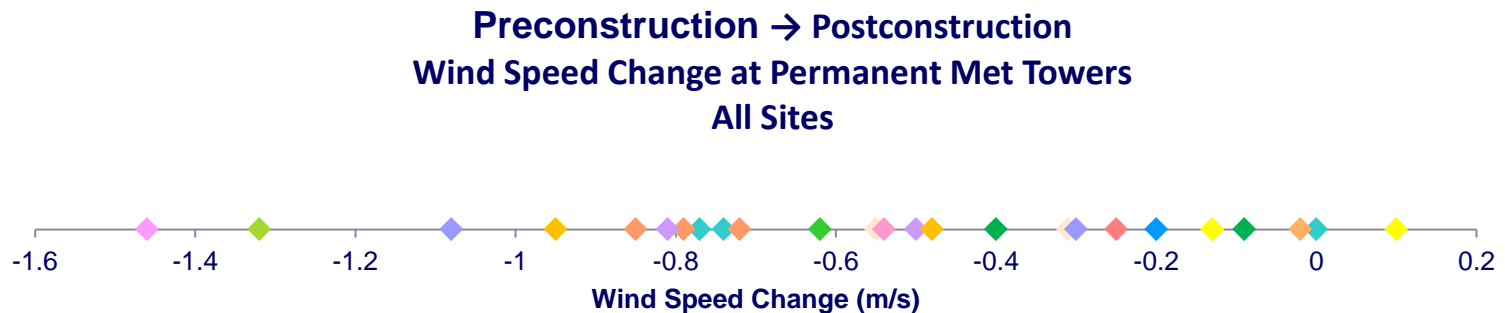
# Pre-Post Build Tower Analysis

- Use data from towers located onsite one year before and after wind farm.
- Use machine learning and MERRA to overlap datasets and calculate difference.
- Limited number of sites (23) and towers (typically one or two per site).



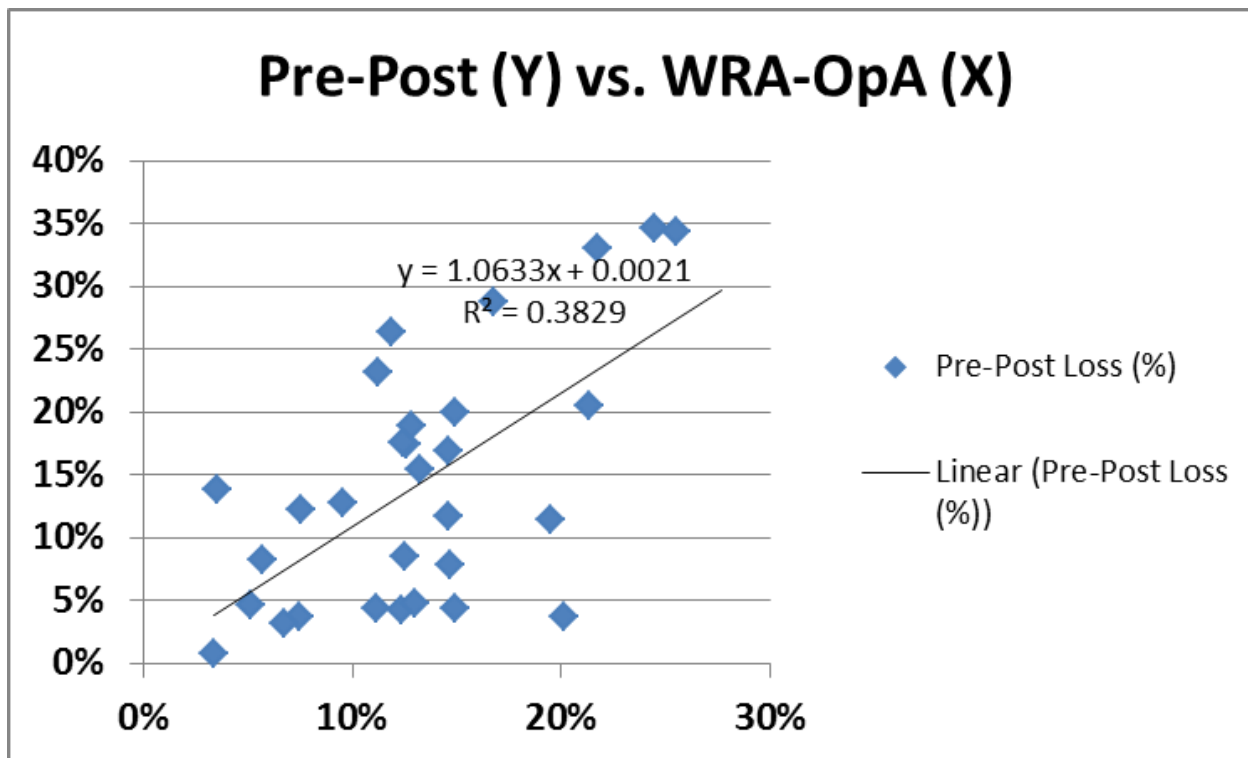
# Pre-Post Build Tower Analysis Results

- The impact of turbine wakes on the analyzed met speeds is substantial
- Speed reductions at the towers range from 0.04 m/s to almost 1.5 m/s
- Impacts were much larger at tower locations in the middle of the farms



# WRA Errors and Resulting WRA-OpA Wake Errors

- Mischaracterization of wake of at least several percent is likely within any single farm
- Over the whole fleet, these errors average to near zero
- Our two sources of wake truth show only a tenuous relationship due to both WRA GCF sensitivity and spatial sampling effects of the pre-post towers

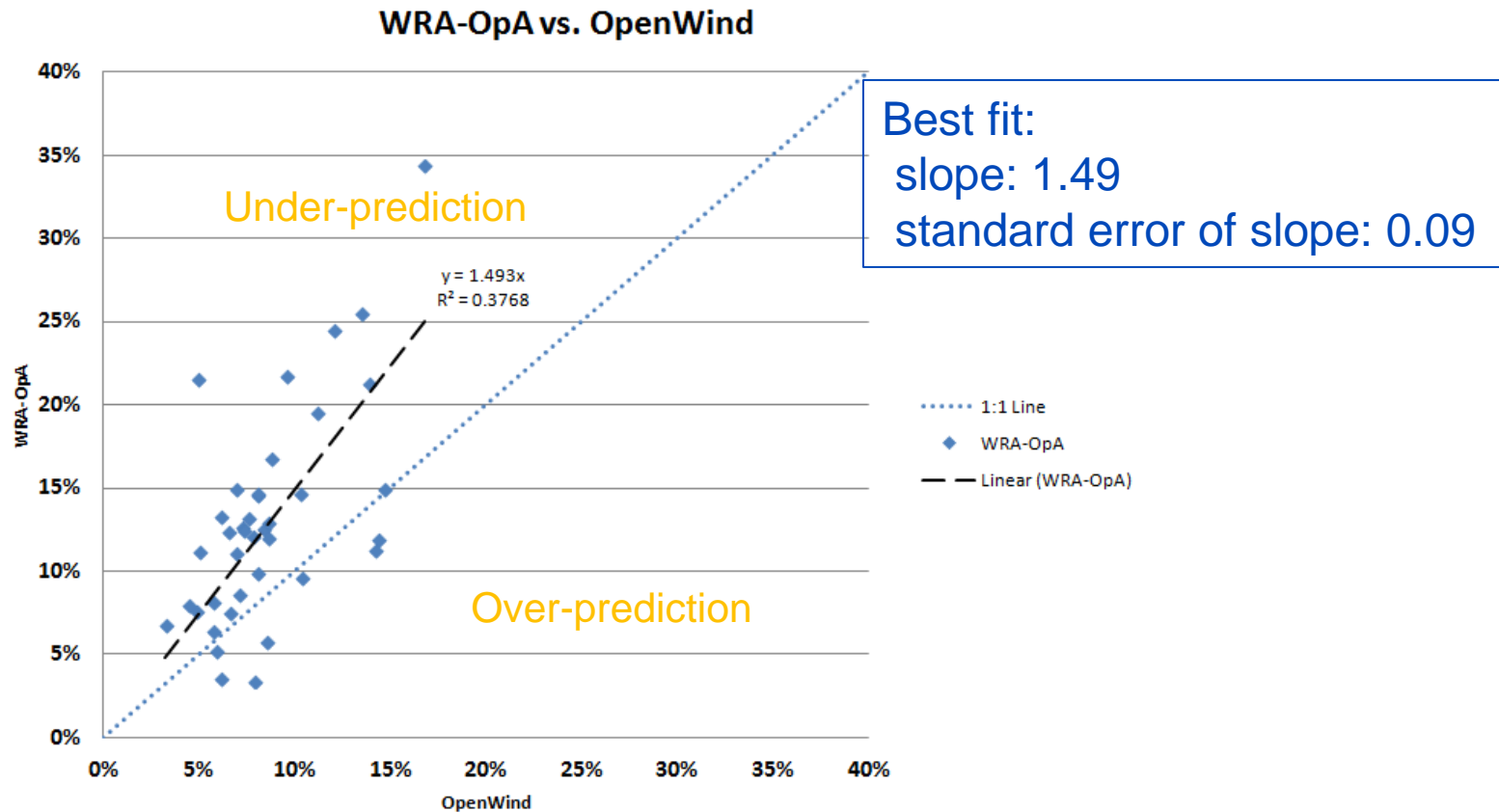


## Two Sources of Pre-Build Wake Estimates

- **Wake models (OpenWind and WindFarmer)**
  - Simplified physics
  - Fast run time
- **Weather Research and Forecast model (WRF)**
  - Complex physics representation
  - Long run time

# OpenWind/Wind Farmer

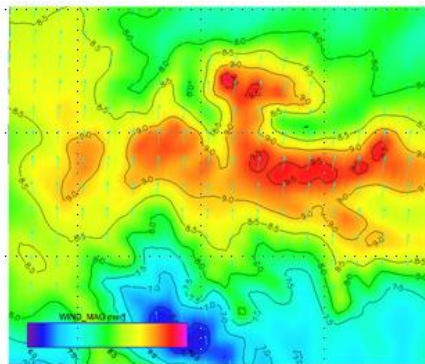
- Legacy versions of these programs were confirmed (from recent literature) to under-predict wakes for large farms
- New versions of both include large wind farm correction (Deep Array), which in acts as an increase in surface roughness.
- Our tests show that these new versions also under-predict wake by ~33% for both small and large farms



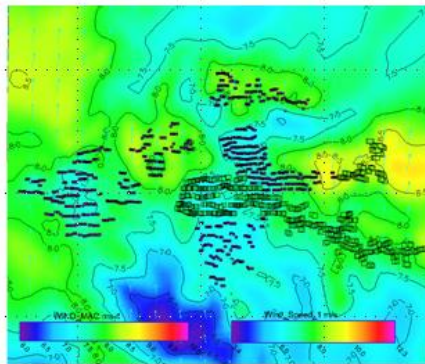
# WRF Wake Simulations

- Gridded model simulation to cover the same year for waked and unwaked runs
- Subtract “turbines” run from “no turbines” run to find wake impacts.
- 28 wind farms have been simulated using WRF

WRF Simulations: Horse Hollow, Callahan, and neighboring wind farms

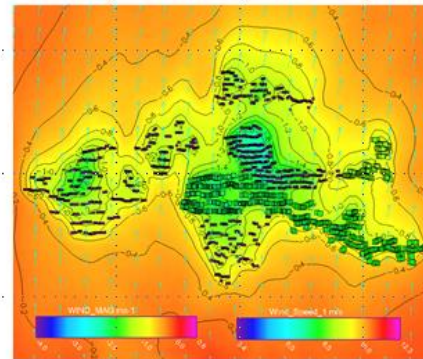
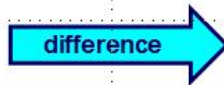


without turbines



with turbines

Because of terrain complications, the only way to separate terrain and waking effects is through the use of numerical simulations both with and without wind turbines.

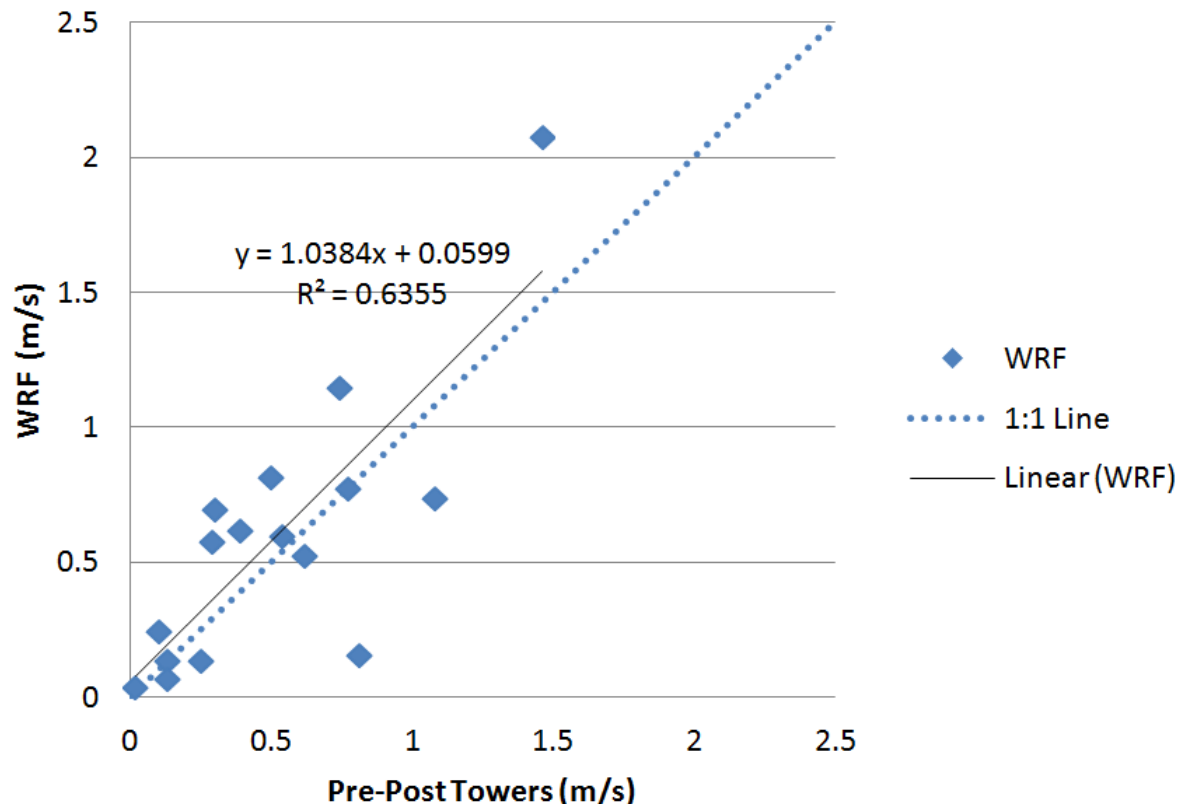




## Intercomparison: WRF and Pre-Post Towers

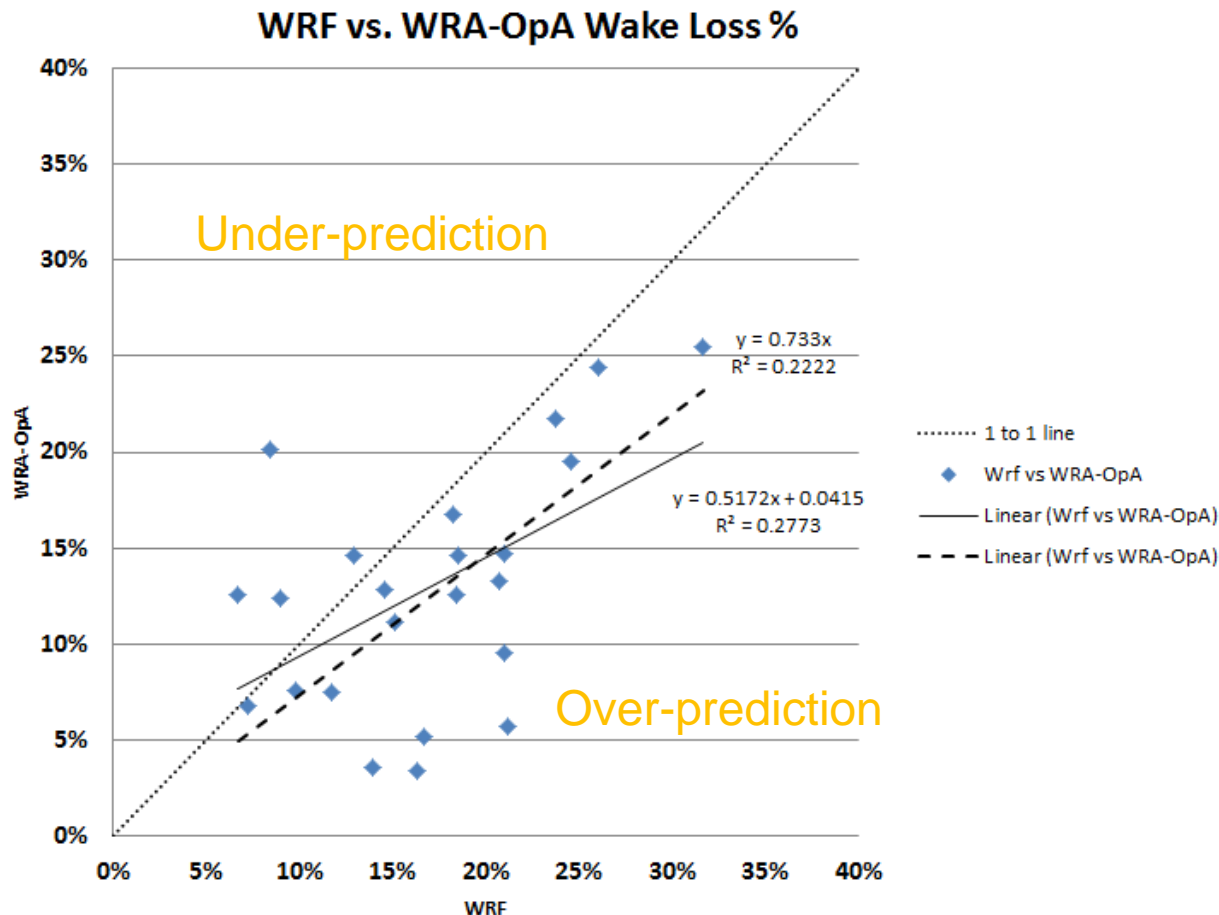
- Continuous spatial coverage of wake speed deficit from WRF allows us to sample at the Pre-Post tower locations for direct comparison against our best wake observations, showing good agreement

WRF vs. Pre-Post Towers Speed Reduction



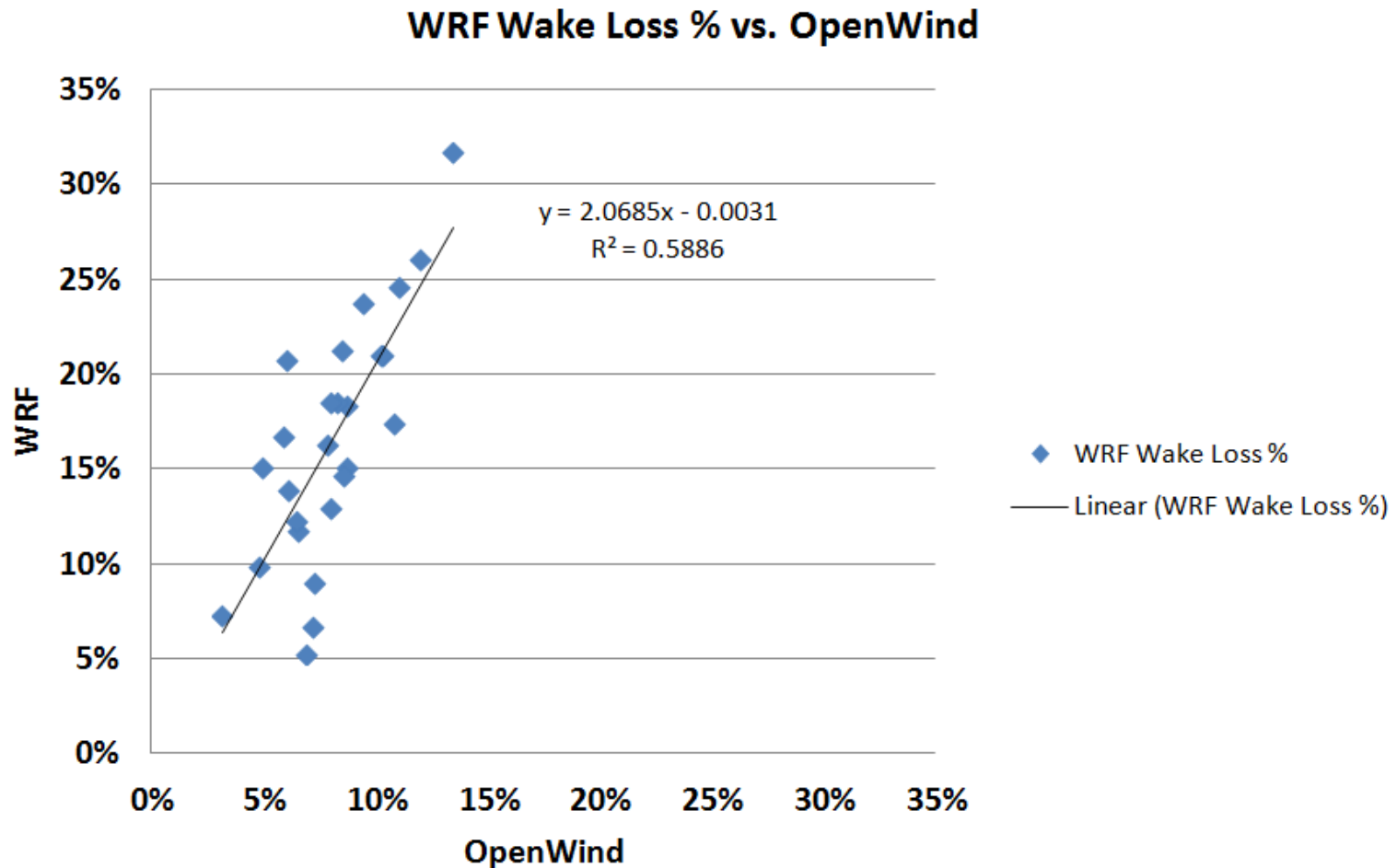
# Intercomparison: WRF and WRA-OpA

- Our tests indicate actual wake 27% less than predicted by WRF
- Because of thrust coefficient curve and WRF high speed bias, WRF over-predicts wake while accurately predicting speed difference.



# Intercomparison: WRF and OpenWind

- OpenWind and WRF correlate well, indicating that the two physics-based approaches are responding to the same relevant site features



## Conclusions

- **Scatter inherent in the WRA-OpA method limits its usefulness as a direct regression target**
- **Best use of the WRA-OpA results is to scale the physics-based results**
- **Given the good correlation between WRF and OpenWind, we can use these deep array wind models and apply a correction factor to use in most projects**
  - WRF represents the reference solution for research and refinement

## Future Work

- **Continue to research options for running WRF-with-turbines sufficiently fast to support operational use in the future (MIC cards, Cloud, time sub-sampling)**
- **Implement more detailed data management campaign that covers pre and post construction time periods**
- **At existing wind farms, exploit novel situations, such as the farm tripping off, with continuity of wind observations, to further infer wake impacts**
- **Continue to participate in wind community-wide wake research efforts**

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## **Appendix**

## References

- *The OpenWind Deep-Array Wake Model: Development and Validation.* Brower and Robinson, AWS Truepower. 2011.
- *Modelling and Measuring Flow and Wind Turbine Wakes in Large Wind Farms Offshore.* Barthelmie et al., Wind Energy. 2009; **12**:431-444.
- *OpenWind Theoretical Basis and Validation.* AWS Truepower. 2010.
- *Wake effects within and between large wind projects: the challenge of scale, density, and neighbours- onshore and offshore.* Phillips et al., Garrad Hassan, Presented at EWEC 2010.
- *Quantifying the Impact of Wind Turbine Wakes on Power Output at Offshore Wind Farms.* Barthelmie et al., *Journal of Atmospheric and Oceanic Technology*, **27** (2010), 1302-1317.
- *Evaluation of wind farm efficiency and wind turbine wakes at the Nysted offshore wind farm.* Barthelmie and Jensen, Wind Energy, **13** (2010): 573-586.

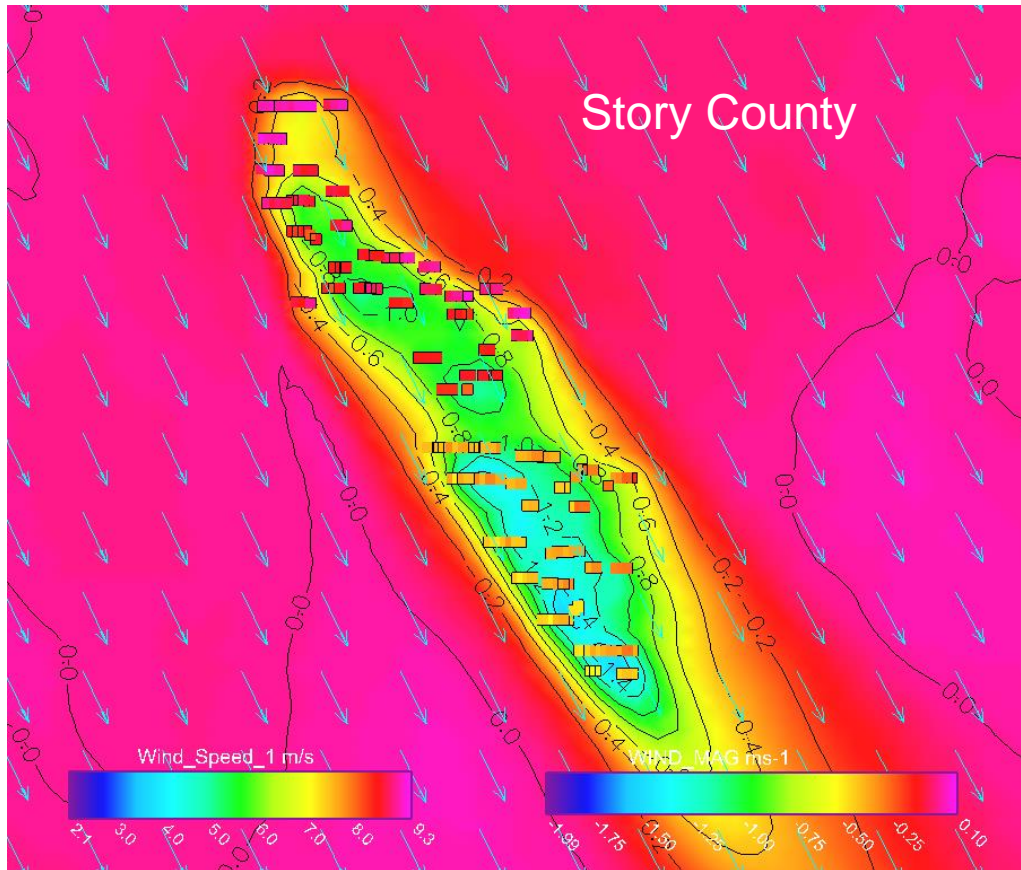
## Active Wake Research Programs

- Iowa State University
- University of Colorado
- WakeBench (IEA Subtask)
- Los Alamos National Laboratory
- University of Wyoming
- Texas Tech University



# WRF Sector Analysis

- Directional sector analysis at the simulated farms confirmed that the pattern of WRF speed difference (turbines minus no turbines) was reasonably accurate



$N = 3360$

