

# Wind turbine blade load characterization under yaw offset at the SWiFT facility



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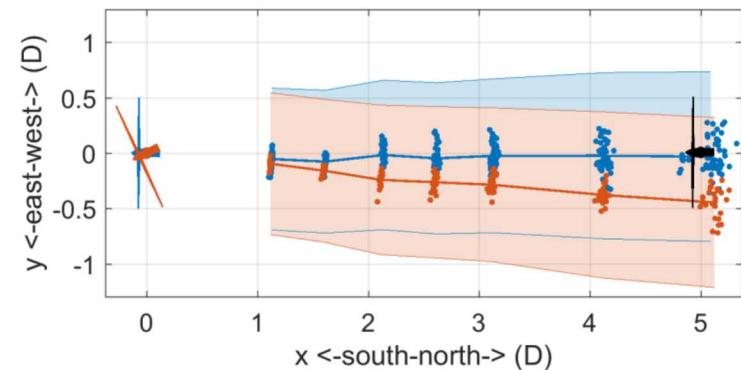
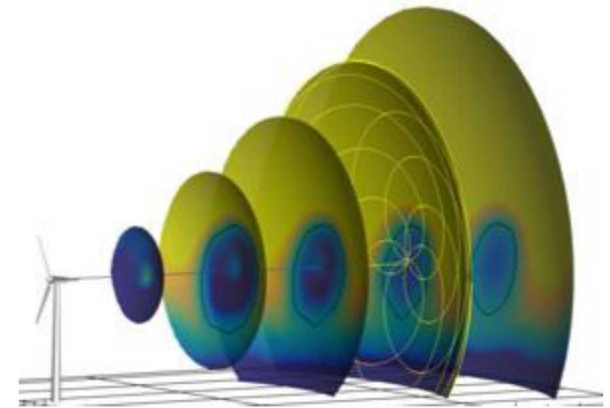
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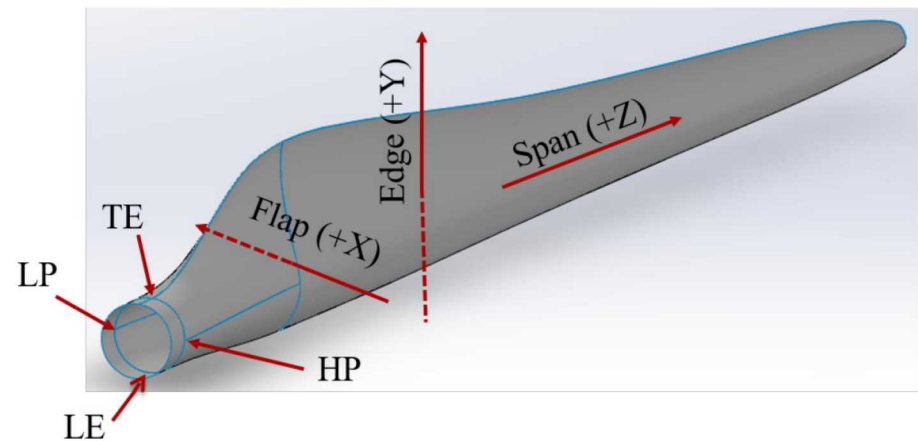
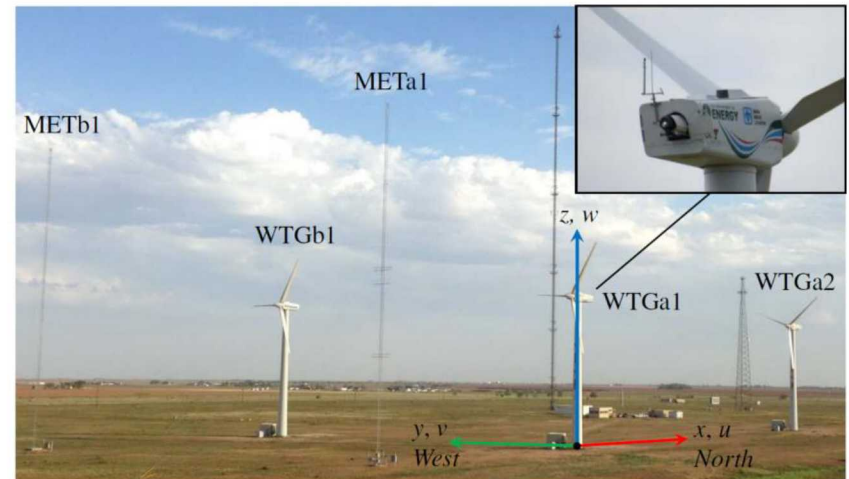
- Wind turbines operate in stochastic flow environments with constantly changing wind speed and direction
- Yaw systems are designed to respond slowly to not over-exercise their components meaning wind turbines nominally operate with an unintentional yaw offset
- There has been recent interest in intentionally yawing wind turbines out of the wind as a wind plant control strategy to improve system power benefits
- For both intentional and unintentional yaw offsets, there is a need for better understanding of the effect on fatigue loads to improve the operation of future wind plants

- Research has been performed on intentionally yawing wind turbines out of the wind at the DOE/SNL Scaled Wind Farm Technologies (SWiFT) facility during the summer of 2017
- The Wake Steering Experiment was performed in collaboration with the National Renewable Energy Laboratory as part of the U.S. Department of Energy's Atmosphere to Electrons (A2e) program
- The experiment includes data from an upstream meteorological tower, two wind turbines spaced 5-diameters apart, and the DTU Spinner Lidar scanning the wake behind the dominantly upstream turbine



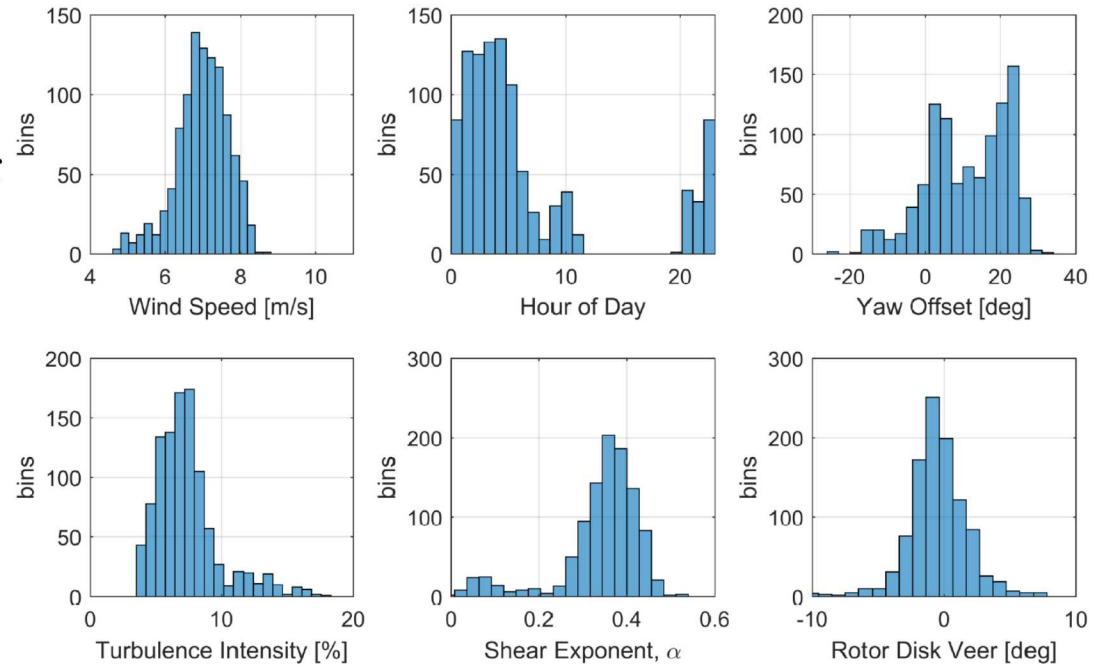


- Data acquired from WTGa1 and METa1 between July 9-13, 2017 are used in this analysis.
- Data are sampled at 50 hz and logged with a site GPS time.
- WTGa1 analyzed data include:
  - optical blade root strain gage pairs calibrated to blade root flapwise and edgewise bending moments
  - generator electrical power output
  - wind turbine azimuth position and rotational speed
  - nacelle heading
- METa1 analyzed data include:
  - sonic anemometer measured wind speed/direction at 5 tower heights





- Data are separated into 50-rotation bins (70-150 sec)
- Data are filtered for a minimum power threshold, operation in Region 2, and for unobstructed wind directions
- Due to the campaign objectives the data are biased towards high shear, low turbulence conditions with positive yaw offset
  - Bulk data trends are biased towards high shear and low turbulence compared to the site averages for the shear exponent of 0.21 and turbulence intensity of 12%







## Rotational Spectral Analysis:

- Operational loads data are processed through a rotational Fourier analysis
- Data are separated into 50-rotation bins and resampled into evenly discretized  $15^\circ$  azimuth positions
- Amplitude and phase frequency content is calculated for each bin at per-revolution (P) forcing's

$$X_k(\omega_\theta) = \sum_{n=1}^N x(\theta) e^{\frac{-i2\pi(k-1)(n-1)}{N}}$$

## Fatigue Calculations:

- Blade load fatigue cycles are extracted through rainflow cycle counting of the first 60-sec from the 50-rotation bin's time series
- Damage equivalent loads (DEL) for the flapwise and edgewise root bending moment are calculated and compared as ratios to a reference DEL
- A fatigue slope of  $m=10$  for fiberglass is used
- The increase in blade damage ( $D$ ) caused by the operational conditions is the normalized DEL raised to the fatigue slope ( $m=10$ )

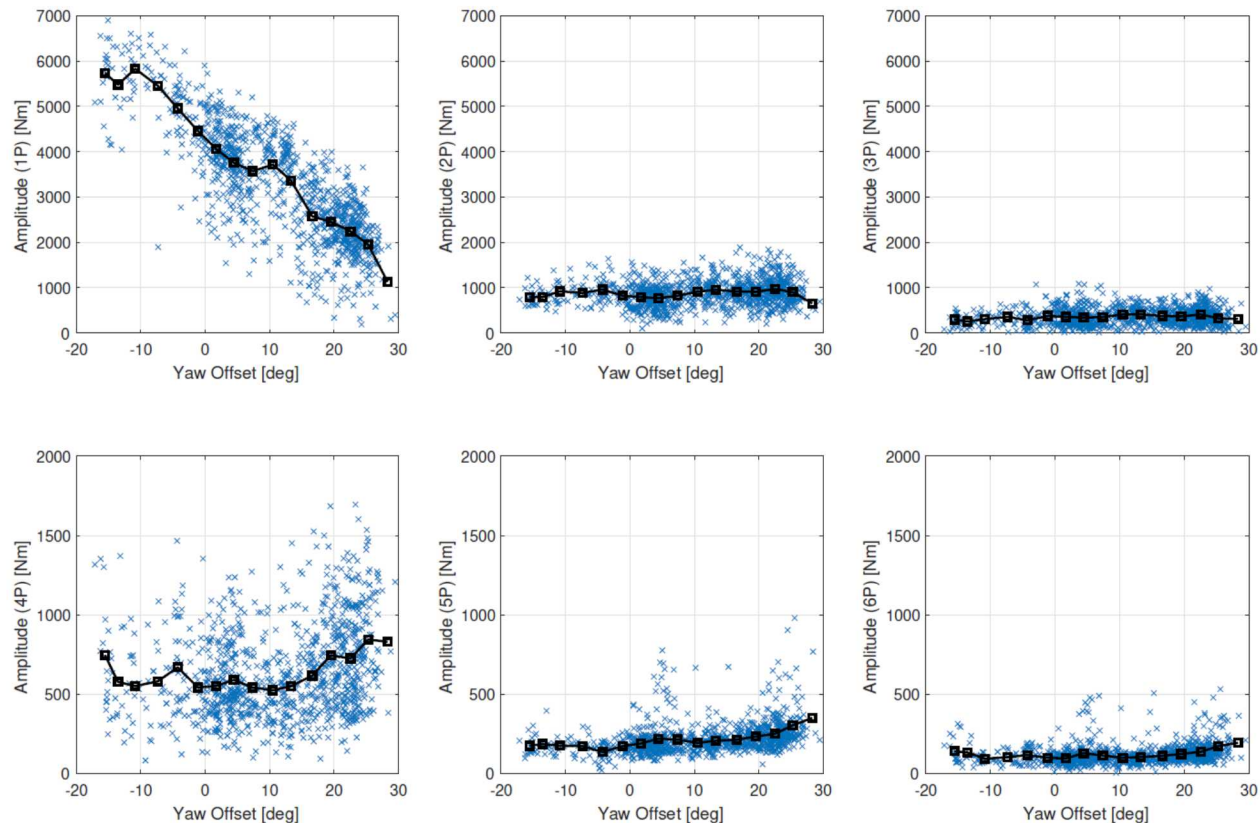
$$M = M_u \cdot N^{-1/m}$$

$$M_e = \frac{M_a}{1 - \frac{M_m}{M_u}}$$

$$D = \sum_{i=1}^k \frac{n_i}{N_i} \Big|_{M_{e,i}}$$

$$DEL = M_u \left( \frac{10^6}{D} \right)^{-1/m}$$

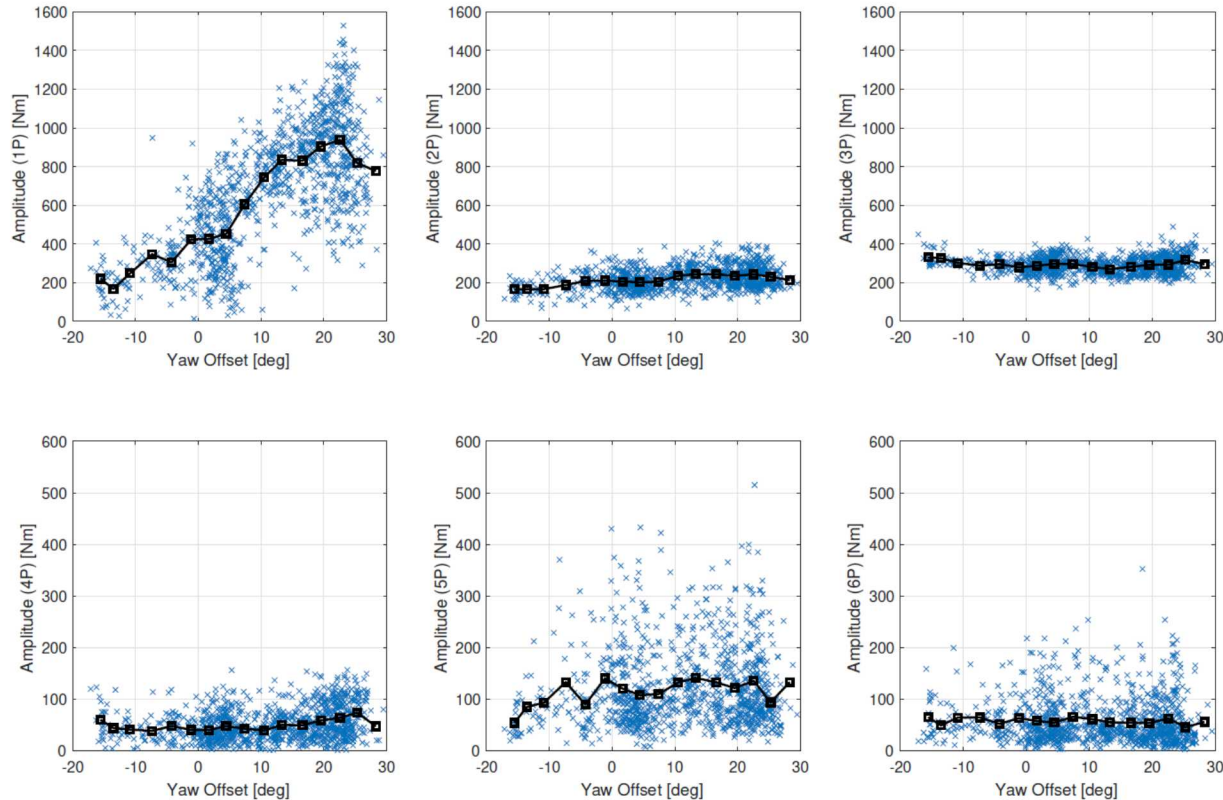
$$\frac{D_2}{D_1} = \left( \frac{DEL_2}{DEL_1} \right)^m$$



(a) Rotational spectral analysis flapwise bending moment amplitude content

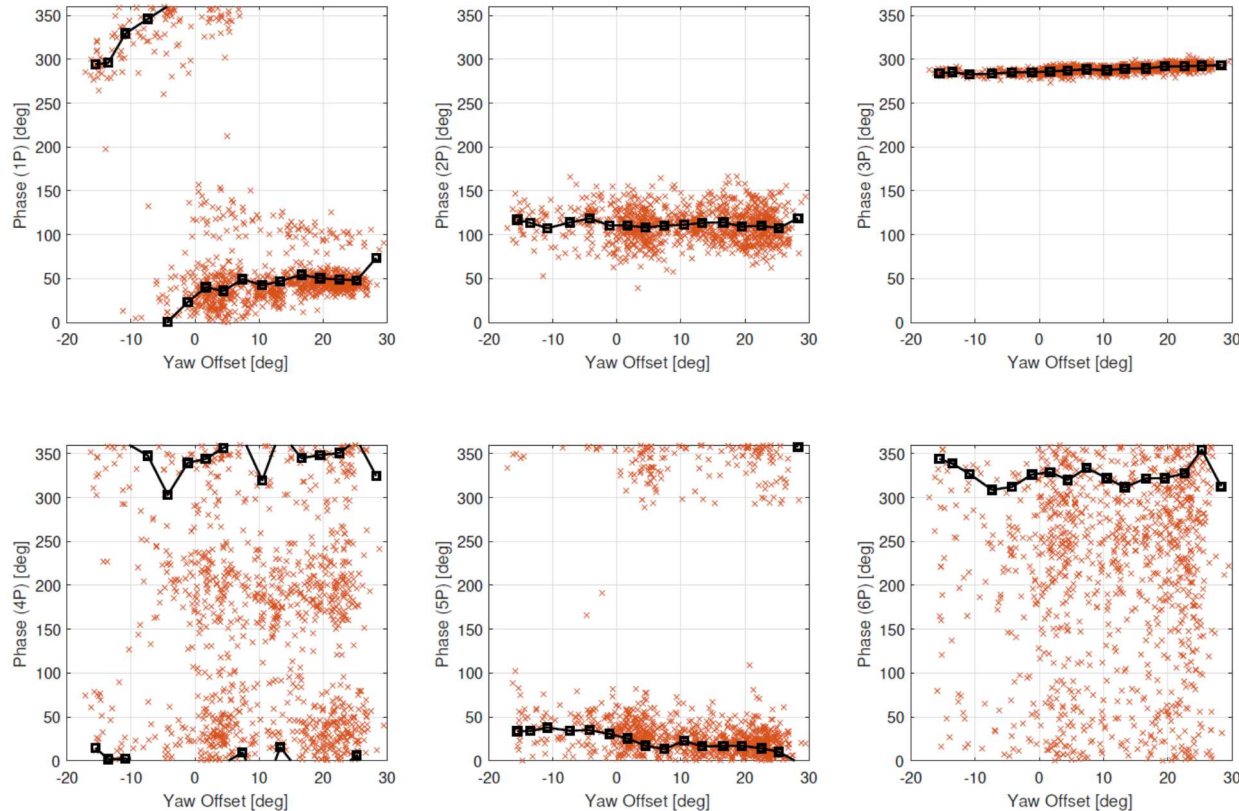
- Strong, negative correlation between 1 per-revolution amplitude and yaw offset for flapwise bending moment
- 2-6 per-revolution amplitude content are uncorrelated with yaw offset





(a) Rotational spectral analysis edgewise bending moment amplitude content

- Strong, positive correlation between 1 per-revolution amplitude and yaw offset for edgewise bending moment
  - The 1P amplitude correlations show opposite trends for flapwise and edgewise bending moments
- 1P amplitude is best signal to track for yaw offset identification

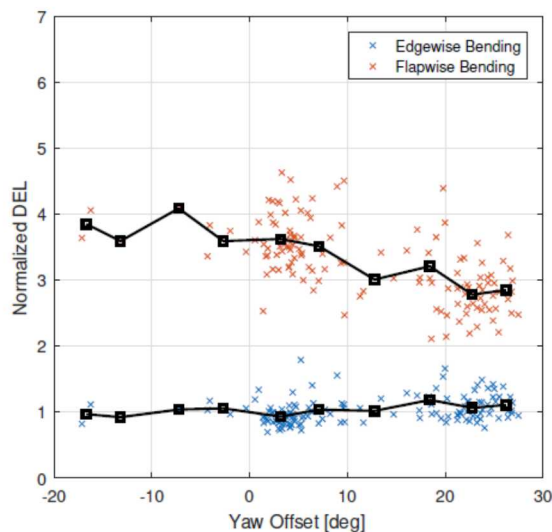


(b) Rotational spectral analysis edgewise bending moment phase content (wrapped from  $0^\circ$  to  $360^\circ$ )

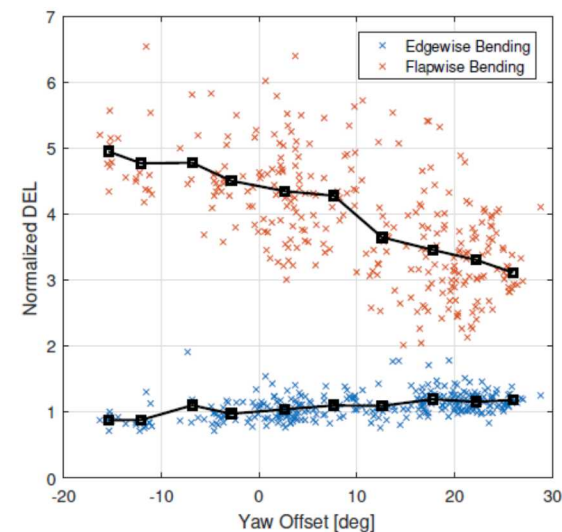
- Phase content for flapwise and edgewise bending moments show weak correlations with yaw offset which typically are within the data's spread
- 1P edgewise phase is most strongly correlated, where the phase angle moves towards the downstream portion of the rotor with yaw offset
- 1P flap- and edgewise phase angle are only equal at a  $-10^\circ$  yaw offset



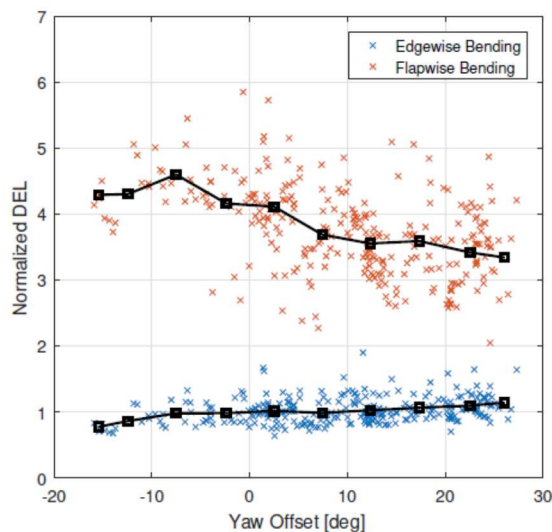
- Damage equivalent loads are calculated for each bin and normalized by the edgewise DEL at zero yaw
- The effect of wind speed on DEL can be more substantial than yaw offset so the data are further grouped into 0.5 m/s bins
  - Bin averages are calculated for every 5° yaw offset bins
    - The lowest wind speed bin (a) has insufficient negative yaw data so its trend for negative yaw should be disregarded



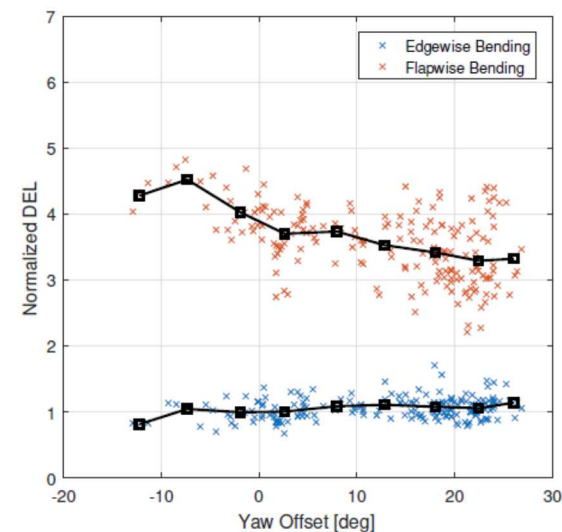
(a) Wind speed between (6, 6.5) m/s



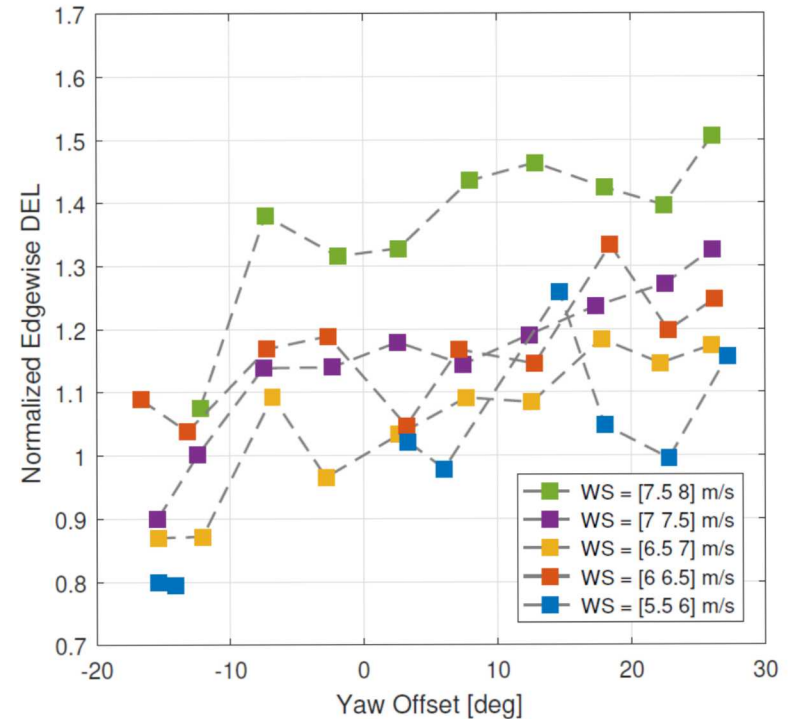
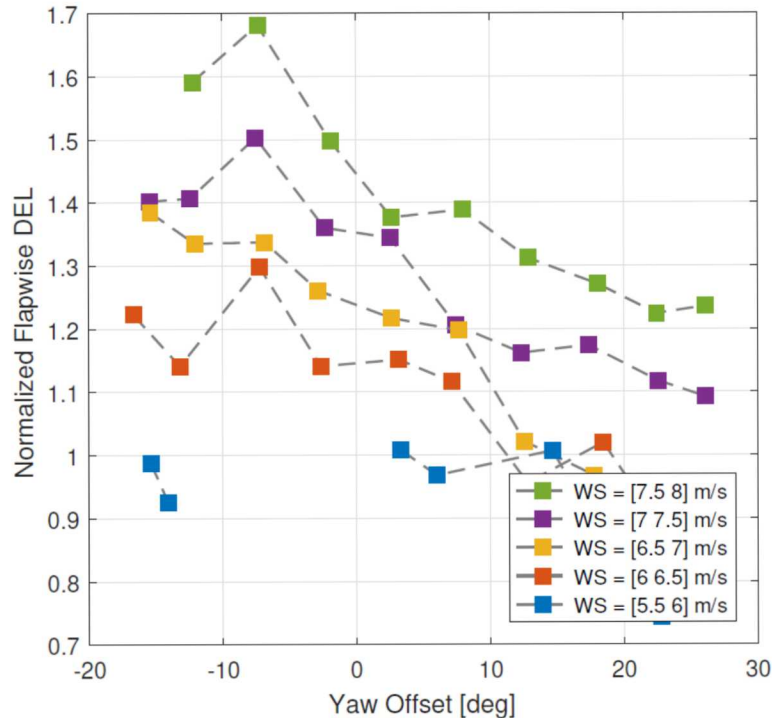
(b) Wind speed between (6.5, 7) m/s



(c) Wind speed between (7, 7.5) m/s

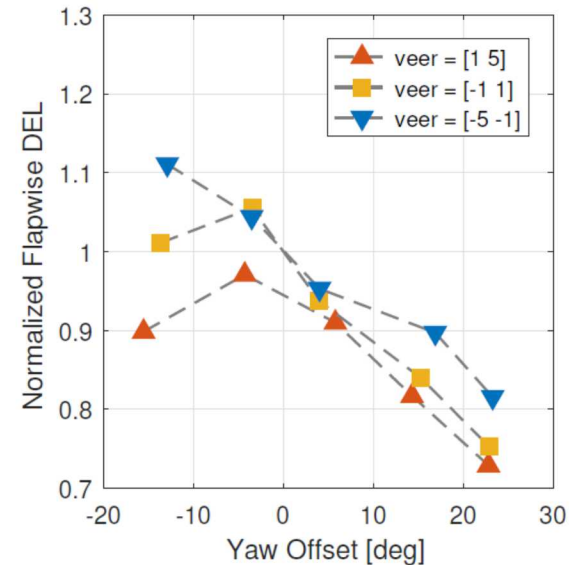
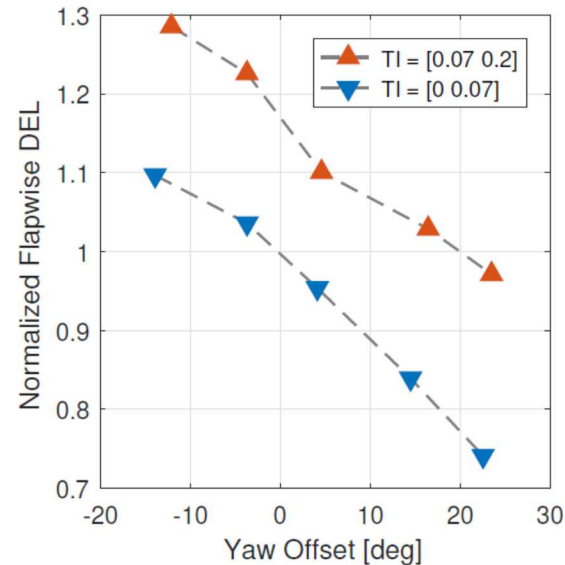
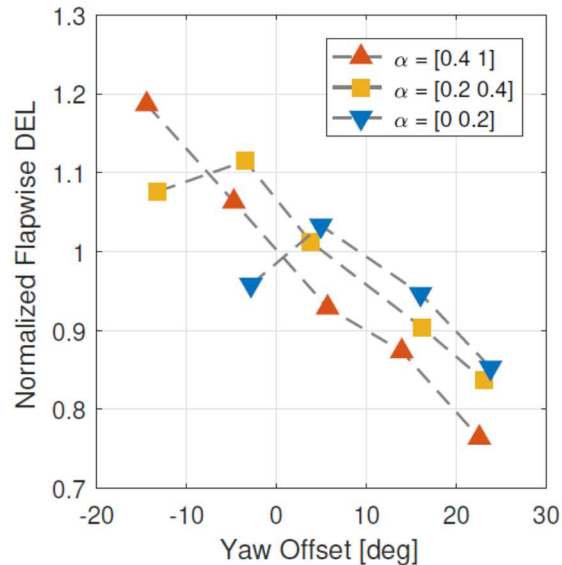


(d) Wind speed between (7.5, 8) m/s



- The correlation of DEL with wind speed is seen to be as substantial as the yaw offset
- Data reveals that flapwise DEL increases with negative yaw and decreases with positive yaw as the velocity shear loading is balanced
- Edgewise DEL shows the negative trend that was observed in the 1P amplitude content, but with a shallower slope





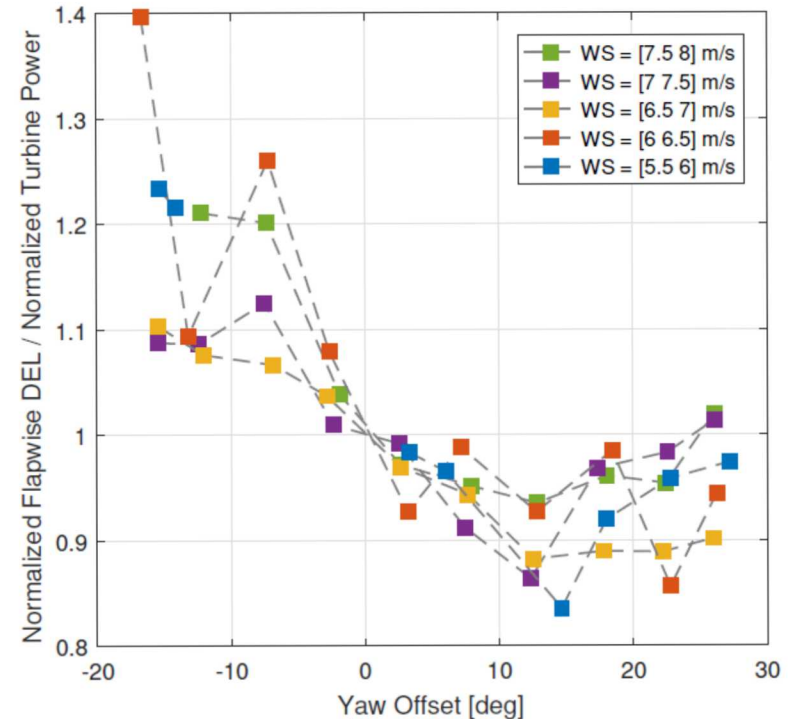
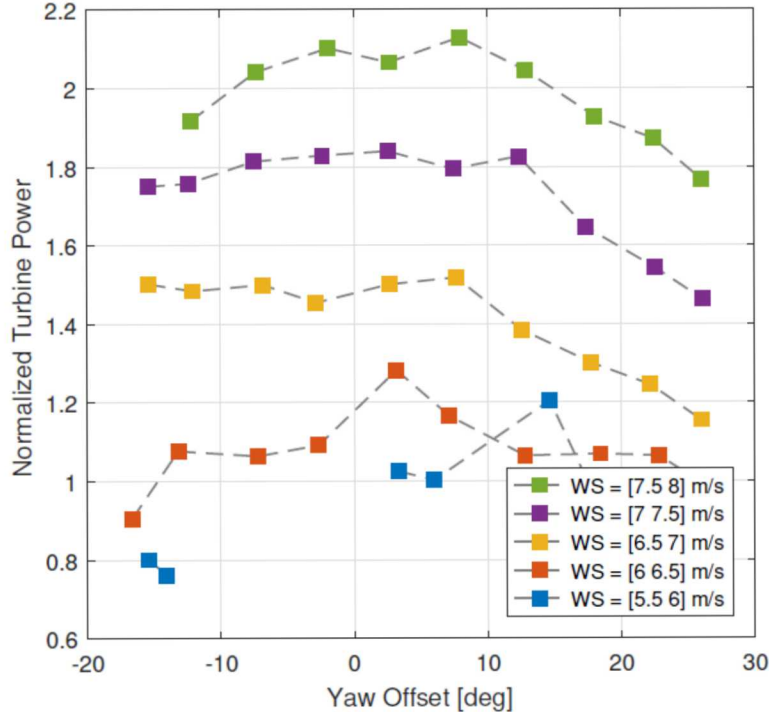
(a) Velocity shear DEL trends

(b) Turbulence intensity DEL trends

(c) Wind veer DEL trends

- The data are further divided into atmospheric states including all of the wind speed data from 6-8 m/s to ensure sufficient data content
- In positive yaw, the DEL for low shear cases decreases at a lower rate than high shear cases, as expected
- Turbulence intensity has a stronger observed effect than shear
- Wind veer is not strongly correlated, although for positive yaw negative veer appears to increase DEL while positive veer decreases DEL





- Yaw offset is observed to reduce power beyond around  $\pm 10^\circ$
- Yaw offset reduces power and alters the fatigue loads for wind turbines, both negatively and positively for the SWiFT turbines
- An effective “cost” is defined which compares DEL and power, normalized to the wind speed bin’s zero yaw offset values
  - For the high-shear data analyzed, the SWiFT turbine has the best overall performance at around  $10\text{-}12^\circ$  yaw offset

## Conclusion



- The Wake Steering experimental data were analyzed to compare blade loads with yaw offset under varying atmospheric conditions
- Flapwise bending DEL from the SWiFT turbine was observed to increase with negative yaw and decrease with positive yaw, based on the level of shear across the rotor disk
- Edgewise bending DEL has the opposite trend with yaw
- The 1P amplitude spectral content is most correlated with yaw offset, is the source of DEL increase/decrease, and may be the best signal to track to identify yaw offset magnitude
- Turbulence intensity has a stronger effect on DEL versus yaw offset than shear was observed to have
- Based on loads reduction and a relatively constant power, wind turbines may have optimal performance at a nominally positive yaw offset, based on the atmospheric conditions

The complete Wake Steering experimental dataset will be available for download at [a2e.energy.gov](https://a2e.energy.gov) by August.



## Opportunities to tour the following facilities:

- The Department of Energy & Sandia National Laboratories Scaled Wind Farm Technology (SWiFT) facility
- Texas Tech University Ka-Band Radar
- Group NIRE Distributed Energy Management
- GE Prototype Wind Turbines
- DNV-GL Lidar certification

*[energyworkshops.sandia.gov](http://energyworkshops.sandia.gov)*