General Response

Thank you for your comments and the time taken to provide this review. Please find in this document a response to your specific comments. The authors are confident the responses provided will clear up any confusion or concerns regarding the methods used. Please find as supplemental material an edited version of the manuscript with changes tracked.

Foremost, the authors would like to state that at the time of the experiment (14 December 2014), experimental validation of wind plant control at full scale was rare. Similarly, there was a dearth of observational data (especially compared to numerical simulation) characterizing in three-dimensions wind turbine wake structure and variability. Therefore, in collaboration with an industry partner, the preliminary objectives of the experiment were to (1) examine three-dimensional wind turbine wake response to changes in wind turbine yaw and blade pitch and (2) examine how these changes impact the net power production of individual turbines in the wind plant (i.e. quantifying the effectiveness of the wind plant control strategy). However, experimental logistics (e.g. experimental control limitations imposed by the wind plant operator) ultimately inhibited execution of these experimental objectives. Therefore, experimental objectives and analysis focus evolved to exploring the complexities and difficulties associated with performing a wind plant control experiment at full scale. The results of this study lend insight into these difficulties, and therefore, should be used to inform future field campaigns. The authors are confident the changes made in response to your comments allow the manuscript to tell a more direct story.

Specific Comments

Pg. 2, ln. 17: Double check your references listed. For example, Vollmer et al. 2016 and Fleming et al., 2018 are listed as wind tunnel experiments, but these are numerical simulations.

Vollmer et al. (2016), Fleming et al. (2018), and also Park and Law (2016) were based on numerical simulation. By mistake, these works were initially miscited in the manuscript as studies based on wind tunnel experiments. Furthermore, although Jiménez et al., (2010) leverages the results of previously performed wind tunnel experiments, a significant portion of the manuscript was based on numerical simulation. The citing of these manuscripts was appropriately modified in the manuscript and the authors are confident that the referenced works are now properly cited.

The text now reads (Pg. 2 Lns. 15 through 20):
“The benefit of wind plant control has been previously demonstrated using numerical simulation (e.g. Jiménez et al., 2010; Johnson and Fritsch, 2012; Lee et al., 2013; Annoni et al., 2015; Park and Law, 2015; Fleming et al., 2015; Gebraad and van Wingerden, 2015; Gebraad et al., 2016; Park and Law, 2016; Vollmer et al., 2016; Fleming et al., 2018; Kanev et al., 2018) and in wind tunnel experiments (e.g. Parkin et al., 2001; Corten and Schaak, 2003; Howland et al., 2016; Schottler et al., 2017; Bartl et al., 2018; Bastankhah and Porté-Agel, 2019).”
Furthermore, the reference for Fleming et al. (2019) has been updated to reflect its status being upgraded from ‘In Review’ to ‘Published’.

Pg. 2, ln. 22: Another recent full-scale validation of wind plant control is: Howland et al, Wind farm power optimization through wake steering, Proceedings of the National Academy of Sciences, 2019.

Thank you for bringing this work to our attention, it is now cited in the manuscript (Pg. 2 Ln 19). Furthermore, based on your input the authors discovered another relevant study by Howland wherein wind tunnel experiments were used to examine the impact of yaw operation on wake structure. This work (documented below) is now referenced in the manuscript.


Pg. 2, ln. 25: “To expand upon existing full-scale validation efforts, agreements were made with an industry partner...” Please be sure to review the existing full-scale validation efforts and explain how the present work fits in.

As mentioned in the manuscript, full-scale validation of wind plant control remains limited. Despite the potential benefit of wind plant control being demonstrated in both numerical simulation and wind tunnel experiments, full-scale validation of these control techniques must be performed before wind plant control can be commercially employed.

Based on your comment, the authors recognize that the phrasing,

“...To expand upon existing full-scale validation efforts, agreements were made with an industry partner to modify the yaw and blade pitch of a utility-scale wind turbine for a limited time period to examine the resulting variations in wake structure.”

is vague when it comes to defining exiting full-scale validation efforts, and also does not denote how the presented research expects to contribute to the research field.

Existing experimental validation of wind plant control at full-scale has nominally focused on either (1) quantifying the benefit of wind plant control by analyzing the power and controls data of individual turbine pairs in a wind plant over extended periods (e.g. Fleming et al., 2017a; Ahmad et al., 2019; van der Hoek et al., 2019; Fleming et al., 2019; Howland et al., 2019) or (2) using advanced measurement technologies (e.g. lidar) to examine near-wake response to turbine control changes (Trujillo et al., 2016; Fleming et al., 2017b).

The referee should consider the date of the experiment (14 December 2014) when factoring in how this research fits into the now currently published research. The initial focus of the manuscript was not to necessarily build on any specific experimental field campaigns (very few experimental studies were published at this time), but rather to contribute to the general dearth of full-scale experimental datasets examining wind plant control and its efficacy. The preliminary objectives
of this experiment were to (1) examine three-dimensional wind turbine wake response to changes in wind turbine yaw and blade pitch in both the near- and far-wake regions and (2) examine how these changes impact the net power production of individual turbines in the wind plant (i.e. quantifying the effectiveness of the wind plant control strategy). However, experimental objectives evolved based on the experimental difficulties encountered. Analysis focus ultimately evolved to exploring the complexities and difficulties associated with performing a wind plant control experiment at full scale.

However, the authors recognize that the principal objectives of this experiment were not properly reflected in the manuscript by simply stating ‘To expand upon existing full-scale validation efforts...’. Therefore, the manuscript text was modified to (1) briefly highlight the state of experimental wind plant control validation at full-scale and (2) to better describe the experimental objectives of the 14 December 2014 experiment. The authors are confident the modified text (copied below) more accurately defines the scope and contents of the manuscript.

Pg. 2 Lns. 24 through 34 and Pg. 3 Lns 1 through 2:

“Experimental validation of wind plant control at full-scale has frequently relied upon the analysis of power and controls data from individual turbine pairs in a wind plant to quantify the benefit of various wind plant control techniques (e.g. Fleming et al., 2017a; Ahmad et al., 2019; van der Hoek et al., 2019; Fleming et al., 2019; Howland et al., 2019). However, few studies have used advanced measurement technologies (such as lidar or radar) to document differences in wake structure due to the turbine control changes implemented as part of wind plant control (e.g. Trujillo et al., 2016; Fleming et al., 2017b). Additionally, these studies almost exclusively limit wake measurement to the near-wake region, and therefore, are unable to monitor the downstream progression of these control-induced wake modifications. To contribute to these full-scale validation efforts, and to expand the downstream extent to which control-induced wake changes are measured, agreements were made with an industry partner to modify the yaw and blade pitch of a utility-scale wind turbine for a limited time period to examine the resulting variations in wake structure. Wake measurements were made using Texas Tech University’s Ka-band (TTUKa) Doppler radars employing dual-Doppler (DD) scanning strategies. However, rather than validating the effectiveness of these wind plant wake-mitigating control strategies, results highlight some of the complexities associated with executing and analysing wind plant control strategies at full-scale using brief experimental periods.”

Section 2.2: Can you explain if the 12 October experiment is at a different site? Are both sites in similar terrain, or are there significant differences between them that should be pointed out.

The 14 December 2014 and 12 October 2015 radar experiments were performed at different sites. This is in part why instrumented tower data was available for the 12 October 2015 SD radar deployment but not for the 14 December 2014 DD radar deployment. However, both sites were located in the US Great Plains and there were no significant terrain differences between the two sites that would impact wind plant complex flow structure or variability. Although confidentiality agreements preclude disclosure of the radar deployment locations, manuscript text was modified
to provide more details regarding the general location (i.e. the US Great Plains) of the two radar deployments.

Pg. 3 Lns. 23 through 24 now reads:
“The technical specifications of the TTUKa radars are further detailed in Table 1 and radar deployment specifics (both performed in the US Great Plains) are provided in the subsections below.”

**Fig. 1: Please explain the meaning of the different colors of wind turbines, including white, to avoid confusion.**

A brief description of the turbines was provided in the initial manuscript submission in the first paragraph of Section 3.

Pg. 5 Lns. 11 through 16 and Pg. 6 Lns. 1 through 2:
“Located in the DD domain of the 14 December 2014 deployment were 20 wind turbines distributed across two turbine rows. The wind turbines were characterized by a hub height of 80 m and a rotor diameter (RD) of 101 m. Supervisory control and data acquisition (SCADA) information detailing the turbine inflow wind speed (subject to the nacelle transfer function [NTF]), turbine yaw orientation, and blade pitch angle were provided at a one-hertz sampling frequency from 14:00:00 UTC to 16:59:45 UTC for seven of the wind turbines (denoted by the non-black circles in Fig. 1). Three of the seven wind turbines were located in the leading row of the wind plant, while the remaining four were located in the trailing row. The three lead-row wind turbines were separated by an average distance of 1512.2 m (~15 RD) from the trailing turbine row and were laterally separated from each other by an average distance of 321.1 m (~3 RD).”

However, the authors recognize that this text is disconnected from the description of the 14 December 2014 radar deployment (i.e. Section 2.1) and Fig. 1. Therefore, this text was moved to the end of Section 2 and was also slightly modified to better explain the figure (in particular the meaning of the different color turbines). Furthermore, both the caption and legend to Fig. 1 were modified to improve reader comprehension. The revised text is copied below for reference.

Pg. 4 Lns. 14 through 23:
“Located in the DD domain of the 14 December 2014 deployment were 20 wind turbines distributed across two turbine rows. The wind turbines were characterized by a hub height of 80 m and a rotor diameter (RD) of 101 m. Supervisory control and data acquisition (SCADA) information detailing the turbine inflow wind speed (subject to the nacelle transfer function [NTF]), turbine yaw orientation, and blade pitch angle were provided at a one-hertz sampling frequency from 14:00:00 UTC to 16:59:45 UTC for seven of the wind turbines (denoted by the non-black circles in Fig. 1). Three of the seven wind turbines were located in the lead row of the wind plant (denoted by the blue, red, and purple circles in Fig. 1), while the remaining four were located in the trailing row (denoted by the white circles in Fig. 1). The three lead-row wind turbines were separated by an average distance of 1512.2 m (~15 RD) from the trailing turbine row and were laterally separated from each other by an average distance of 321.1 m (~3 RD). The wake of these
three lead-row wind turbines (referred to as the TL, TT, and TR) were analyzed to examine the effectiveness of the implemented wake-mitigating control strategies.”

**Pg. 6, ln. 1:** What purpose do the downstream turbines (white circles) serve in this experiment.

SCADA data from the four downstream turbines were used along with SCADA data from the three lead-row wind turbines to estimate the region three pitch schedule used in analyses. This was indicated in Sect. 3.1.1 of the initial manuscript submission.

**Pg. 7 Lns 13 through 16:**
“Data from all seven wind turbines were used to ensure a robust estimate of the region three pitch schedule (a total of 41,343 SCADA wind speed and blade pitch angle measurements were used); measurements inconsistent with region three pitch operation and TT data from the experimental periods were not considered when constructing the pitch schedule.”

However, to improve reader comprehension, the manuscript text was slightly modified to that copied below.

**Pg. 7 Lns 29 through 32:**
“Data from all seven wind turbines with SCADA information provided (i.e. the non-black circles in Fig. 1) were used to ensure a robust estimate of the region three pitch schedule (a total of 41,343 SCADA wind speed and blade pitch angle measurements were used); measurements inconsistent with region three pitch operation and TT data from the experimental periods were not considered when constructing the pitch schedule.”

**Pg. 6, ln. 18:** Fleming et al., 2018 deals with numerical simulations, do you mean 2019?

Thank you for noticing this, the text was appropriately modified to Fleming et al. (2019).

**Pg. 7, ln. 6:** If the benefit of modifying blade pitch is greater in region 2, then why was the sole half-hour experiment period in region 3? Would have it made more sense to wait for more favorable conditions?

The wind plant control experiment would have been ideally performed for extended experimental durations and in an environment more conducive to the effectiveness of the implemented control strategies. However, experimental logistics ultimately inhibited execution of the experiment when conditions were optimal. These logistical difficulties are detailed below.

Foremost, although the TTUKa radars can comprehensively document wind plant complex flow structure and variability at high spatial resolutions, data availability is dependent on the atmospheric conditions present (namely the size distribution of scatterers and aerosols present in the ABL). This dependency is also true for other remote sensing instrument such as lidar. Given the specifications of the TTUKa radars, data availability is enhanced in certain precipitating
environments. Therefore, wind plant control experiments were limited to periods when the TTUKa radars were expected to have sufficient data availability. Further impacting rapid radar deployment when atmospheric conditions were ideal was the relative proximity of the wind plant to the staging (i.e. storage) location of the radars. The wind plant was located several hundred miles away from the TTUKa radars when not in use. Therefore, accurate forecasts of atmospheric conditions conducive to data availability had to be made with sufficient lead time to (1) enable proper coordination with the wind plant operators (control experiments were not permitted at all times) and (2) to deploy the radars to the wind plant site. Successful execution of the wind plant control experiment was therefore not as simple as waiting for conditions to become ideal. Furthermore, periods conducive to data availability were not necessarily correlated with atmospheric conditions conducive to the effective implementation of wind plant control (as demonstrated in the manuscript).

However, the 14 December 2014 deployment was not the only attempt at performing a wind plant wake mitigation control experiment. Wind plant control experiments were also performed at a wind plant elsewhere in the US Great Plains and at the TTU Scaled Wind Farm Technology (SWiFT) facility (albeit this is a scaled wind farm facility). The 14 December 2014 radar deployment was presented in this manuscript in part because the wind plant operator allowed a relatively wide range of experimental turbine control changes to be implemented. Prior to recent years, some wind farm operators were reluctant to collaborate on these types of experiments for fear that implementing wind plant wake-mitigating control techniques might impact turbine warranty.

Finally, this field experiment, albeit performed in somewhat of a suboptimal environment and having inherent limitations, lends insight into the complexities associated with performing a full-scale wind plant control experiment. The authors are confident this manuscript adds value to the scientific community and can be used to inform future field campaigns.

Pg. 7, ln. 9: “To maintain the rated generator speed in region three, the wind turbine follows a pitch schedule to extract the desired amount of momentum at various wind speeds.” Blade pitch controllers typically use generator speed feedback to control blade pitch to regulate generator speed. Therefore, if you are adding a pitch offset in region three, what else are you changing in the controller so that the pitch controller doesn’t simply compensate for the offset to bring the generator speed back to rated? Is the generator torque or generator speed setpoint also changed? Could it be that the pitch offset that is added is simply an offset to the “fine pitch” (minimum pitch) angle that the turbine operates at below rated, and that there is no real change to the pitch control above rated? More detail about the intended pitch offset strategy would be helpful.

Due to the proprietary nature of the information, the construct of the wind turbine controller was not provided by the turbine manufacturer. Therefore, the authors are unable to expand on the blade pitch angle offset strategy, nor are they able to state with confidence why these changes were not effectively implemented. This is reflected in the manuscript on Pg. 7 Lns. 3 through 5:

“Not having access to wind turbine controller design is a major challenge to fully understanding wind turbine behaviour (Fleming et al., 2019), or rather, how the controller
responds to variable inflow conditions when attempting to enact the desired control offsets. Therefore, the provided discussions do not detail why the turbine was able or unable to enact the desired control changes, but rather focuses on quantifying the resulting offsets.”

However, it was the author’s understanding that the blade pitch angle offsets (i.e. +1°, +2°, +3°) were implemented independent of the inflow state and operation of the turbine (i.e. region two or three).

The referee indicates that generator speed (i.e. rpm) is typically used to help regulate wind turbine blade pitch. Therefore, even if blade pitch were directly modified by incorporating the blade pitch angle offset, the turbine might recognize inconsistent generator speeds relative to the turbine inflow wind speed and appropriately modify the blade pitch. However, regardless of whether generator speed was relied upon to help regulate blade pitch, experimental mean blade pitch angle offsets were not negligible (i.e. Figs. 3 and 4 of the manuscript). Experimental period (i.e. +1°, +2°, +3°) generator speed behavior was also examined to try and gain a more comprehensive understanding of the implemented control changes. However, analyses did not lend much insight, and therefore, these analyses were not incorporated into the manuscript.

Pg. 7, Lns. 29 through 32:
“Data from all seven wind turbines with SCADA information provided (i.e. the non-black circles in Fig. 1) were used to ensure a robust estimate of the region three pitch schedule (a total of 41,343 SCADA wind speed and blade pitch angle measurements were used); measurements inconsistent with region three pitch operation and T_T data from the experimental periods were not considered when constructing the pitch schedule.”

The constructed region three blade pitch schedule is provided below (note: due to confidentiality agreements the blade pitch angle values could not be plotted along the y-axis). In this figure, there is a rough linear relationship between blade pitch and the region three turbine inflow wind speeds. However, detracting from this linear relationship is a small magnitude of clustering in blade pitch between 12 and 13 m s^{-1}. This clustering is believed to occur as a result of a rapid increase (i.e. at timescales less than the response time of the blade pitch controller) in the turbine inflow wind speed from velocities consistent with region two to region three. Within this period, the turbine is unable to appropriately modify its blade pitch angle to be consistent with region three pitch operation. While it would be desirable to remove these measurements, these values are surrounded by blade pitch angle measurements consistent with region three pitch operation and only represent 1.19 % of the total number of measurements considered. Therefore, this clustering is not expected to significantly impact the construction of the region three pitch schedule.
The authors are confident that these methods enable a robust best-estimate of the region three pitch schedule. Further confidence can be placed on these methods based on the results of Duncan et al. (2019).


In Duncan et al. (2019), this pitch schedule was used to estimate variations in generator speed due to suboptimal blade pitch angle activity.

Pg. 9, ln. 12: A 1.45 km by 1.8 km averaging area seems too large for determining the local inflow wind direction to the turbines, especially if you are trying to distinguish between the wind inflow to each of the three turbines. Furthermore, given the advection time across the 1.8 km analysis area, the estimated wind directions are likely not very well correlated with what the turbines see at a high temporal resolution. Can you try this with 100 m x 100m averaging areas, local to each turbine? This could improve your results, or at least make things more meaningful.

The referee contends a 1.45 km by 1.8 km averaging area is too large to discern turbine-specific inflow differences. Therefore, the referee argues the value of \( \theta_{\text{inf}} \) will likely not be well correlated with the turbine inflow wind direction at higher temporal resolutions. In order to more accurately resolve turbine inflow conditions, the referee recommends using a 100 m by 100 m averaging area local to each turbine. However, due to the interpretation of the DD synthesized wind fields, this 100 m by 100 m turbine-local averaging area is also not necessarily appropriate. Denoted on Pg. 3
Ln. 30 and Pg. 12 Ln 9, the TTUKa radars take on average 60.4 s to collect a single DD volume, and within this DD volume acquisition period, turbulent structures move. Therefore, denoted in Sect. 2.1, Pg. 4 Lns. 5 through 7:

“The DD wind maps can be interpreted as a pseudo-average of the wind conditions over the volume acquisition period, where the DD volume time stamp denotes the end of the volume period.”

Space-to-time conversion techniques (e.g. Duncan et al., 2019) are required to accurately resolve temporal inflow variability at sub-volume timescales (i.e. less than the measurement revisit period). Using local measurements without a proper understanding of how turbulent structures move within the DD volume acquisition period can confound results. Ideally, space-to-time conversions would have been used in Sect. 3.2 to develop a semi-continuous stream of turbine inflow information (wind speed and direction) over the DD volume acquisition period. This would have allowed for temporal inflow variability to more accurately defined, and therefore, would have enabled a more comprehensive determination of the implemented control changes. Denoted on Pg. 12 Lns. 11 through 16:

“To improve controller assessment, future field campaigns should place precedence on turbine inflow measurements (wind speed and direction) independent of the turbine control system (i.e. non-SCADA data). Experiments using scanning-based measurements should use advanced analysis techniques, such as those established in Duncan et al. (2019) wherein space-to-time conversions were performed on the spatially distributed velocity fields, to provide a comprehensive characterization of the turbine inflow wind speed and direction on a second-by-second basis. Application of these methods was limited because of data availability issues.”

These methods were not employed because of data availability issues. While a more local turbine inflow estimate (specifically laterally) might have slightly improved result accuracy, the use of a larger-scale averaging area does not reduce the meaningfulness of the results. Given the average time between DD volume scans (i.e. 60.4 s) and the mean wind speed within the DD analysis period (i.e. 13.71 m s\(^{-1}\) as defined by the mean wind field within the freestream analysis area), a relatively large averaging area is not a bad estimate of the mean conditions observed at the turbine over the DD volume acquisition period. While the freestream analysis area is still larger than the longitudinal wind field area that might advect through the turbine over the DD volume acquisition period (based on Taylor’s hypothesis this would be a distance of 828.1 m [i.e. 60.4 s x 13.71 m s\(^{-1}\)]), differences between these two inflow estimates are not expected to be significant.

Regardless, analysis was performed to demonstrate the sensitivity of the turbine inflow estimate to the averaging area used. Comparisons were made between the turbine inflow wind direction determined using (1) the freestream analysis (i.e. \(\theta_{\text{inf}}^V\)) and (2) local turbine averaging areas representing the estimated portion of the upstream wind field that will advect through the turbine over the subsequent volume acquisition period based on Taylor’s hypothesis. A comparison of these different inflow estimates at 15:31:00 UTC is demonstrated in the figure below. The left subplot demonstrates the spatial dimensions of the individual averaging areas at 15:31:00 UTC, while the right subplot demonstrates the turbine inflow wind direction profile derived from these
averaging areas. The profile-averaged turbine inflow wind direction is provided in the legend of the right subplot for each averaging area. Despite analyzing significantly different areas (especially laterally), differences between the turbine inflow wind direction estimates were small. At 15:31:00 UTC, the $T_L$ inflow wind direction was 0.7° greater than the value of $\theta_{inf}$, the $T_T$ inflow wind direction was equal to the value of $\theta_{inf}$, and the $T_R$ turbine inflow wind direction was 1.2° less than the value of $\theta_{inf}$.

The same analysis was performed for each DD volume in the analysis period to demonstrate that these small differences were not rare. Time histories of $\theta_{inf}$ and the local turbine inflow estimates are provided in the figures below for each turbine. The black lines denote the $\theta_{inf}$ time history and the colored lines denote the local turbine inflow estimates. The local turbine averaging areas were redefined in each DD volume using Taylor’s hypothesis based on the mean conditions in the freestream analysis area. While Taylor’s hypothesis was used to give a general idea of the sensitivity of the turbine inflow estimate to the averaging area used, the referee should refer to Duncan et al. (2019) for why Taylor’s hypothesis cannot be systematically relied upon to accurately denote the advection of the upstream wind field and hence why these techniques were not used in the manuscript.

Provided in the legend of each figure are both the mean deviation and the mean absolute deviation between $\theta_{inf}$ and the local turbine inflow estimates for the entire DD analysis period. Time history data gaps are due to data availability issues; at least 50% of the respective averaging area was required to determine the mean wind direction estimate. Considering inflow estimate data from all turbines and DD volumes, the mean deviation between the two turbine inflow estimates (the local turbine inflow estimate minus the value $\theta_{inf}$) was -0.2° and the mean absolute deviation was 1.1°. These results do not indicate that the use of the freestream analysis area to estimate the turbine
inflow wind direction (i.e. $\theta_{\text{inf}}^V$) reduced the meaningfulness of the results. However, the authors do acknowledge that analyses would have benefited from more accurate turbine inflow estimates.
Furthermore, the use of the freestream analysis area to estimate the turbine inflow wind direction provides a common reference frame for defining both the wake (i.e. $\theta_{\text{skew}}^V$) and streak (i.e. $\theta_{S-\text{skew}}^V$) skew angles.

Section 3.1.2: Given the large positive mean yaw errors with or without the offset applied, it seems possible that the yaw position reported in the SCADA data is not calibrated properly. It is common for yaw position values from SCADA data to deviate significantly from the true orientation (i.e., 0 degrees -> true north) over time. If not, this should be discussed further.

The referee raises the concern that yaw calibration errors might have impacted results because improper yaw calibration is not uncommon. Although not discussed in the manuscript, analysis was performed to discern whether a yaw calibration error was present in either the $T_L$, $T_T$, or $T_R$.

Foremost, regardless of any yaw calibration error, variations in yaw error (i.e. $\theta_{\text{err}}^V$) should elicit distinct changes in wake skew (i.e. $\theta_{\text{skew}}^V$). Counterclockwise rotor rotation relative to a fixed turbine inflow wind direction (i.e. a net positive increase in $\theta_{\text{err}}^V$) should enhance wake deflection to the right when looking downstream (i.e. resulting in a net increase in $\theta_{\text{skew}}^V$), and clockwise rotor rotation relative to a fixed turbine inflow wind direction (i.e. a net negative decrease in $\theta_{\text{err}}^V$) should enhance wake deflection to the left when looking downstream (i.e. resulting in a net decrease in $\theta_{\text{skew}}^V$). However, demonstrated in Fig. 10b and the figure below, it is tough to discern any distinct trends in $\theta_{\text{skew}}^V$ resulting from variations in $\theta_{\text{err}}^V$. Based on the figure below, it appears as though the value of $\theta_{\text{skew}}^V$ slightly decreased as the value of $\theta_{\text{err}}^V$ increased. However, when data from all three turbines is considered (i.e. Fig. 10b), and thereby a wider range of $\theta_{\text{err}}^V$ values considered, this relationship was not as apparent. Therefore, yaw calibration errors cannot explain the general insensitivity of in $\theta_{\text{skew}}^V$ to variations in $\theta_{\text{err}}^V$. 

![Graph showing inflow wind direction estimate over time](image-url)
Alternatively, assume each turbine has a yaw calibration error and therefore is generally well aligned with the inflow wind direction in the DD analysis period (i.e. each turbine exhibits a mean $\theta_{\text{err}}$ value of 0°). If this is true, then the wake of each turbine should roughly extend linearly downstream an angle consistent with the value of $\theta_{\text{inf}}$. To examine whether the wind turbine wakes were extending linearly downstream, the $T_L$, $T_T$, and $T_R$ wake orientation angles (i.e. $\theta_{\text{wake}}$) were compared to value of $\theta_{\text{inf}}$ in the figure below.

The left subplot demonstrates the values of $\theta_{\text{wake}}$ were consistently less than the volume-respective value of $\theta_{\text{inf}}$. Therefore, the wake centerlines were on average skewed to the left of their expected location based on the value of $\theta_{\text{inf}}$ and an assumed $\theta_{\text{err}}$ value of 0°. The right subplot demonstrates the difference between the $T_L$, $T_T$, and $T_R$ $\theta_{\text{wake}}$ values and the value of $\theta_{\text{inf}}$. On average, the $T_L$ $\theta_{\text{wake}}$ value was offset of $\theta_{\text{inf}}$ by -2.90°, the $T_T$ $\theta_{\text{wake}}$ value was offset of $\theta_{\text{inf}}$ by -3.34°, and the $T_R$ $\theta_{\text{wake}}$ value was offset of $\theta_{\text{inf}}$ by -2.17°. Negative values indicate wake skew to the left when looking downstream. Because the values of $\theta_{\text{wake}}$ were inconsistent with the volume-respective value of $\theta_{\text{inf}}$, and also because the $\theta_{\text{skew}}$ values demonstrated little sensitivity to variations in $\theta_{\text{err}}$, yaw calibration errors were not expected to have a significant impact on the presented results.
To provide the reader with more confidence that yaw calibration errors did not adversely impact the results, text was added to the manuscript.

Pg. 17 Lns. 2 through 4

“Furthermore, the values of $\theta_{\text{skew}}^V$ were inconsistent with $\theta_{\text{inf}}^V$, which along with the weak correlation between $\theta_{\text{skew}}^V$ and $\theta_{\text{err}}^V$ provides confidence that the unexpected mean sign of $\theta_{\text{skew}}^V$ was not simply a result of a turbine yaw calibration error.”

Pg. 11, ln. 15: ‘However, nacelle-based measurements are inherently distorted...’ Another factor to consider is that the flow distortion from the rotor can change as the control changes. Adding a pitch offset could cause the wind speed behind the rotor to change differently than with the original control, complicating the detection of changes in turbine operation as a function of wind speed.

The referee indicates that experimental turbine control changes could impact the validity of the nacelle-transfer function (NTF). These turbine control changes could modify typical flow distortion levels behind the rotor, thereby reducing the effectiveness of the NTF used to convert the nacelle-mounted anemometer measurement to a rotor-effective inflow velocity. Therefore, experimental control changes could also impact detection of the control changes implemented when analyzed as a function of the NTF-based inflow velocity. As denoted prior, this demonstrates the need for more accurate turbine inflow estimates, in particular measurements that are independent of the turbine control system. Manuscript text was slightly modified to reflect the need for these types of measurements.

Pg. 12 Lns. 11 through 16:

“To improve controller assessment, future field campaigns should place precedence on turbine inflow measurements (wind speed and direction) independent of the turbine control system (i.e. non-SCADA data). Experiments using scanning-based measurements should use advanced analysis techniques, such as those established in Duncan et al. (2019) wherein space-to-time conversions were performed on the spatially distributed velocity fields, to provide a comprehensive characterization of the turbine inflow wind speed and direction on a second-by-second basis. Application of these methods was limited because of data availability issues.”

Pg. 12: ln. 10: ‘Both of these factors might have contributed to the experimental control offsets not being fully realized.’ Certainly for yaw control, a single 10-minute period might not be sufficient to observe meaningful yaw misalignment changes, given the slow dynamics of yaw controllers.

Additional text was added to the manuscript to better convey this message.

Pg. 12 Lns. 25 through 27:

For example, the brief duration of the experimental periods may have been insufficient to realize/validate the desired control offsets (in particular the ability to observe significant yaw misalignment changes), and the ability to implement the desired control
changes might have been impacted by the ABL conditions present (e.g. region three inflow wind speeds).

Section 4.1: I would suggest a revised wake tracking algorithm in light of improvements in the understanding of wake deflection physics. As discussed in papers such as the following, yaw misalignment can cause wakes to have a ‘curled’ shape due to the presence of counter rotating vortices. This means the peak velocity deficit could change with height and averaging across all heights in the rotor disk area is not necessarily the most relevant metric.

- Fleming et al., A simulation study demonstrating the importance of large-scale trailing vortices in wake steering, Wind Energy Science, 2018.

A more meaningful lateral wake center estimate for wind plant control applications can be found using the method explained in Vollmer et al. 2016, where the cubed wind speed is averaged across a hypothetical rotor disk area centered at different lateral displacements. The displacement that results in the lowest value can be considered the wake center position.

The referee contends the employed wake-tracking algorithm might not have adequately resolved lateral wake center location because of the impact yaw misalignment can have on wake structure. However, before addressing this concern, the authors would like to clarify the methods used in the wake-tracking algorithm. The referee states ‘averaging across all heights in the rotor disk area is not necessarily the most relevant metric’; however, no averaging was used in the wake-tracking algorithm. Instead, at each constant-height plane within the wind turbine rotor sweep, the horizontal location of the wind speed minimum was determined and the median (not mean) horizontal location of these minima was used to discern the lateral wake center location.

Regardless, an analysis was performed to examine whether the operational yaw errors modified wake shape. Both numerical simulation and wind tunnel testing demonstrate rotor-sweep relative variations in wind turbine thrust caused by yaw error can promote a kidney-shaped (or curled) wake (e.g. Churchfield et al., 2016; Vollmer et al., 2016; Bartl et al., 2016). Therefore, wake structure changes induced via yaw error could potentially impact the wake-tracking algorithm. To briefly examine changes in wake shape due to yaw error, composite mean wake deficit cross-sections were developed for the T_L, T_T, and T_R using data from all contributing DD volumes in the +10° experimental period.

To determine the composite mean wake deficit cross-section, the vertical wake cross-sections used to identify the wake were first redefined at each distance downstream about the wake center location. Wind speed measurements within the redefined cross-sections were then converted to wake deficit values, defined as the percentage reduction in wind speed from the DD volume freestream wind speed profile. The DD volume freestream wind speed profile was developed by
determining the mean wind speed within the freestream analysis area at each constant-height plane within the wind turbine rotor sweep (i.e. the same as Fig. 5 but for wind speed instead of wind direction). Therefore, the individual wake deficit values (i.e. \(v_{\text{deficit}}\)) were defined by,

\[
v_{\text{deficit}} = \frac{V_{\text{wake}, h} - V_{\text{inf}, h}}{V_{\text{inf}, h}} \times 100
\]

where \(V_{\text{wake}, h}\) denotes a wake cross-section wind speed measurement at height \((h)\) and \(V_{\text{inf}, h}\) denotes the height-respective DD volume freestream wind speed. The wake deficit cross-sections were then averaged at each distance downstream to determine the TL, TT, and TR composite mean wake deficit cross-sections for the +10° experimental period. Wake deficit contours at 10 %, 15 %, and 20 % were used to qualitatively examine wake shape. However, based on this contour analysis at 1 RD downstream (refer to the figure below), neither the wake of the TL, TT, nor TR exhibited a modified (i.e. kidney or curled) shape. The contours are roughly concentric, and thereby denote a near circular wake. The most notable difference between the composite mean wake of the TL, TT, and TR was the relative size of the contours (i.e. indicating differences in wake intensity).

**Reference Figure:** Composite mean wake deficit cross-sections (presented as a percentage reduction from the DD volume freestream wind speed profile) at 1 RD downstream for the (a) TL, the (b) TT, and (c) the TR. Contours corresponding to a 10 %, 15 %, and 20 % wake deficit are overlaid in black and the wind turbine rotor sweep is overlaid in red. (d) The TL, the TT, and the TR 15 % wake deficit contour at 1 RD downstream.

While this analysis was limited to the +10° experimental period, additional analysis was performed using data from all DD volumes within the DD analysis period (i.e. 14:22:32 UTC – 15:31:57 UTC). At 1 RD, 2 RD, and 5 RD downstream, the mean horizontal (i.e. lateral) location of the wind speed minima (defined relative to the lateral location of the wake center [i.e. +/-]) was determined at each constant-height plant within the wind turbine rotor sweep for the TL, TT, and TR. The mean lateral location of these wind speed minima is denoted in the figures below. Variability about the zero line denotes the average displacement of the height-respective wind speed minima from the wake center. Therefore, changes in wake structure due to yaw error should
be reflected in these mean profiles. Although there were some lateral displacements in the wind speed minima mean location at 1 RD downstream, no mean displacement exceeded 10 m. The observed displacements were not significant enough to indicate extensive wake shape deformation resulting yaw error. These results do not indicate yaw error does not modify wake shape, but suggest the operational yaw error values along with the turbine inflow wind speeds were not significant enough to elicit these changes. Therefore, the effectiveness of the wake-tracking algorithm is not expected to have been adversely impacted by any perceived changes in wake shape resulting from yaw error.

To ease referee concerns regarding the employed wake-tracking algorithm, the referee-recommended wake-tracking algorithm was also applied and analyses was re-ran. The lateral wake center location was determined at each distance downstream by identifying the lateral location that minimized the rotor-sweep disk averaged cubic wind speed. Lateral displacements ranging from -100 m to +100 m at 10-m intervals were examined at each distance downstream. Using this method (refer to the figure below), the T₇ wake at 14:33:20 UTC was located -10 m (i.e. to the left when looking downstream) of its expected location based on the inflow wind direction (i.e. θₗ₉). Despite implementing the referee-recommended wake-tracking algorithm, the wake skew values were not significantly modified. With this wake-tracking algorithm applied, all three turbines still exhibited a negative mean wake deflection angle, indicating wake deflection to the left when looking downstream. This is opposite of expected based on the operational yaw error values and theory. Therefore, the choice of wake-tracking algorithm did not impact the take-home message of the wake skew analysis. In order to maintain consistency between the SD and DD wake-tracking
algorithms, the wake-tracking algorithm previously established by Hirth and Schroeder (2013) will be employed.

**Fig. 9**: Is Fig. 9 (a) showing the distance from the centerline of the wake after corrected for the skew angle, or from the centerline in the mean wind direction. Please clarify what is being shown.

The figure indicates the lateral distance from the wake centerline as defined by $\theta_{\text{wake}}^V$ (i.e. “after corrected for the skew angle”). This is denoted in the first manuscript submission on Pg. 15 Lns. 4 through 5:

> “Constrained linear least-squares regression (Gill et al., 1981) was used to determine the value of $\theta_{\text{wake}}^V$, wherein the wake centerline was required to emanate from the location of the wind turbine.”

To improve clarity, both manuscript and figure text were modified.

Pg. 15 Ln. 13 and Pg. 16 Lns. 1 through 2:

> “The value of $\theta_{\text{wake}}^V$ minimized the error sum of squares; the error distribution was defined as the lateral distance between individual wake center locations and the wake centerline (i.e. $\theta_{\text{wake}}^V$) (e.g. Fig. 9a).”

Figure 9 Caption:

**Figure 9.** (a) The lateral distance between the $T_T$ wake centerline (as defined by $\theta_{\text{wake}}^V$) and the wake center locations (i.e. the $T_T$ error distribution). (b) TTUKa DD hub-height wind speed (m s$^{-1}$) at 15:30:00 UTC overlaid by the $T_L$, $T_T$, and $T_R$ wake centerline and the wake center locations at 1-RD increments.

Pg. 16, Ln 6: ‘...indicating the observed wake deflection...was opposite of that expected.’ How might wind veer impact the wake deflection during the experiment period? Could this be an explanation for the unexpected skew?

Wind veer can induce rotor-sweep relative variations in yaw error. In the presence of wind veer, a rotor that is well aligned with the wind at hub height will exhibit increasing magnitudes of yaw error (albeit opposite signs) in the upper and lower halves of the wind turbine rotor sweep. Therefore, a turbine exhibiting minimal yaw error at hub height should produce a wake (as viewed from above) that widens with distance downstream. This perceived wake widening is due to the wake being deflected in one direction (left/right) in the lower half of the rotor sweep and in the other direction (right/left) in the upper half of the rotor sweep. Depending on the veer profile (i.e. veering/backing and intensity [° per m AGL]), wake deflection can either be promoted or inhibited relative to that expected based on the hub-height turbine inflow wind direction and the yaw orientation angle. For this reason, hub-height wind directions cannot comprehensively
quantify the rotor-relative variations in yaw error that contribute to wake deflection. To increase the robustness of the results, a rotor-sweep average wind direction (i.e. $\theta_{\text{inf}}^\nu$) was used to quantify yaw error. This was noted in the first manuscript submission on Pg. 9 Lns 16 through 18 and Pg. 10 Lns 1 through 5.

“Due to wind plant and turbine measurement limitations, yaw error is traditionally defined relative to the hub-height wind direction measured by the nacelle wind vane. However, the nacelle wind vane is unable to account for differences in wind direction with height (i.e. wind veer). Therefore, yaw error defined by the hub-height wind direction will be unable to comprehensively quantify the rotor-sweep relative variations in the axial induction factor that cause wake deflection. Hence, yaw error (i.e. $\theta^\nu_{\text{err}}$) was defined in each DD volume relative to the RSA average turbine inflow wind direction using

$$\theta_{\text{err}}^\nu = \theta_{\text{inf}}^\nu - \theta_{\text{yaw}}^\nu$$

where $\theta_{\text{yaw}}^\nu$ was the DD volume yaw angle (i.e. the mean yaw angle during the DD volume acquisition period). Positive magnitudes of $\theta_{\text{err}}^\nu$ denote counterclockwise rotor rotation relative to $\theta_{\text{inf}}^\nu$.”

Prior to examining wind veer in the DD analysis period, it can be noted that the $T_L$, $T_T$, and $T_R$ wake center analysis performed at 1 RD, 2 RD, and 5 RD downstream (provided in a previous response) did not exhibit the opposing wake deflection signs in the upper- and lower-halves of the rotor sweep that would be expected if significant wind veer was present. To further demonstrate wind veer was not the cause of the unexpected wake skew angles, wind veer was quantified in each DD volume by analyzing the freestream wind direction profile (i.e. Fig. 5). Both the DD analysis period mean and one-sigma value of these freestream wind directions at each constant-height plane within the vertical depth of the wind turbine rotor sweep are provided in the figure below along with a time history of wind veer (defined as the freestream wind direction difference between 130 m and 30 m) in the DD analysis period. Within the DD analysis period, wind veer ranged from -4.16° per 100 m to +2.22° per 100 m about a mean value of -0.60° per 100 m. Over 80 % of the DD volumes exhibited wind veer magnitudes less than 2° per 100 m. Therefore, although wind veer can impact wake deflection, the wind veer present in the DD analysis period was not significant enough to adversely impact the results of the wake skew analysis.
Pg. 18, ln 3: ‘9 degrees counterclockwise...of $\theta_{\text{inf}}^V$. Stating what $\theta_{\text{inf}}^V$ is would clear up any confusion about the sign convention.

The value of $\theta_{\text{inf}}^V$ was defined in the first manuscript submission on Pg. 9 Lns. 13 through 16:

“The freestream wind direction was defined at each DD constant-height plane within the vertical depth of the wind turbine rotor sweep (Fig. 5b), and the mean of these freestream wind direction measurements was used to determine the rotor sweep area (RSA) average turbine inflow wind direction (i.e. $\theta_{\text{inf}}^V$).”

However, because there is a gap in the usage of this term, manuscript text was modified to refer the reader back to Sect. 3.1.2 for the definition of $\theta_{\text{inf}}^V$.

Pg. 19 Lns. 4 through 5:

“At 15:21:08 UTC, a $\theta_{\text{streak}}^V$ value of 162.72° was determined, which was 9° counterclockwise (i.e. to the left when looking downstream) of $\theta_{\text{inf}}^V$ (defined in Sect. 3.1.2).”
Section 4.2.1: The potential impact of streak orientation on wake skew is an interesting idea. However, a deeper discussion of how this might cause the skewing of the wake would be appreciated.

In the first manuscript submission (Pg. 17, Lns. 4 through 5), it was denoted that previous research demonstrates boundary layer streak orientation can slightly deviate (i.e. clockwise or counterclockwise) from the ABL wind direction. While Lorsolo et al. (2005) is listed as the primary reference for this research, both Morrison et al. (2005) and Foster (2005) have also identified streak orientation angles that deviate from the ABL wind direction. Despite this empirical research, it is unknown why boundary layer streak orientation differs from the ABL wind direction. Rather than presenting and validating hypotheses for the formation of boundary layer streaks, previous research has commonly focused on simply quantifying the spatial and temporal structure of these boundary layer heterogeneities. However, to provide the reader with more background on this research area, the references for Morrison et al. (2005) and Foster (2005) are now provided in the manuscript.

Because the physical mechanisms that govern boundary layer streak orientation are relatively unknown, it is difficult to hypothesize how these streaks directly govern downstream wake progression. However, because the dimensions of boundary layer streaks scale well with wind turbine wakes, it can be hypothesized that the same boundary layer forcings causing streak orientation to deviate from the ABL wind direction might also promote similar deviations in wake orientation. To convey this to the reader, manuscript text was modified.

Pg. 18 Lns. 6 through 7:
“Although it is not fully understood what causes streak orientation to differ from the ABL wind direction, it can be hypothesized that the same boundary layer forcings also promote downstream wake deviation from the ABL wind direction.”

Comprehensively quantifying how boundary layer streaks modulate downstream wake progression is outside the scope of this manuscript. However, the authors hope that establishing this preliminary connection will stimulate future research that examines how boundary layer heterogeneities impact the effectiveness of various wind plant control methods.

Fig. 13: Consider showing the joint probability density of streak skew angle and wake skew angle. This would support your idea of a correlation between the two better.

Thank you for your input. A joint probability density of $\theta_{S\text{-skew}}^{V}$ versus $\theta_{skew}^{V}$ is now provided in the manuscript (i.e. Fig. 14). Corresponding to Fig. 14, the following text was incorporated into the manuscript.

Pg. 19 Lns. 11 through 15:
“Similarities between the $\theta_{S\text{-skew}}^{V}$ and the $T_{L}$, $T_{T}$, and $T_{R}$ $\theta_{skew}^{V}$ distributions are further demonstrated by examining their joint probability density. In Fig. 14 the $T_{L}$, $T_{T}$, and $T_{R}$ $\theta_{skew}^{V}$ distributions were combined and compared to the $\theta_{S\text{-skew}}^{V}$ distribution; darker shades of blue indicate increasing probability of the respective $\theta_{S\text{-skew}}^{V}$ and $\theta_{skew}^{V}$ values. Despite positive values of $\theta_{S\text{-skew}}^{V}$, a large percentage of the skew angles (both $\theta_{S\text{-skew}}^{V}$ and $\theta_{skew}^{V}$)
exhibited counterclockwise rotation relative to $\theta_{\text{inf}}$ (indicating downstream deflection to the left).”

Figure 14. The joint probability density between the $\theta_{\text{S-skew}}^V$ and $\theta_{\text{skew}}^V$ distributions. The $T_1$, $T_R$, and $T_T$ $\theta_{\text{skew}}^V$ distributions were combined to produce this heatmap, wherein darker shades of blue indicate an increasing probability of occurrence for the respective $\theta_{\text{S-skew}}^V$ and $\theta_{\text{skew}}^V$ values.

Section 5.1: SD WTA: For a section title, consider spelling out the acronym.

The acronym WTA is now spelled out in the section title.

Section 5: In addition to wake length and wake meandering, what differences have you observed in the relative magnitude of the velocity deficits for stable vs unstable conditions? This would be a valuable addition to the paper.

The authors agree that this would be a valuable addition to the paper. However, analysis and comparison of the convective and stable ABL wake deficit profiles was hindered by several factors. First, there were significant differences in the turbine inflow wind speed between the convective and stable ABLs (reflected in Fig. 15 [now Fig. 16] using 10-min average data). Although 10-min average wind speeds were provided in the manuscript, high-temporal resolution wind speed data spanning the SD analysis period is provided in the figure below to demonstrate these wind speed differences. The red and blue shaded regions denote the convective and stable portions of the SD acquisition period, respectively. The amount of momentum extracted from the inflow, and therefore the wake deficits immediately downstream, inversely vary as a function of the turbine inflow wind speed. Therefore, it would be difficult to exclusively attribute difference
in the convective and stable wake deficit profiles to changes in ABL stability, rather than simply differences in the turbine inflow wind speed.

A second confounding factor to wake deficit analysis was the variation in measurement height with range (approximately 17.5 m per km moving away from the radar [noted in the manuscript on Pg. 5 Lns. 10 through 11]). This impacted the potential for wake deficit analysis as follows:

1) Wind shear varies between the convective and stable ABLs. In the stable ABL, wind shear is typically larger than in the convective ABL. Therefore, the turbine inflow wind speeds will be inherently larger than the wake velocities because of their relative measurement range and height. A shear-correction would be required to resolve relevant wake deficit profiles.

2) Because measurement height varies with range, different portions of the wake are examined at incremental distances downstream.

For these reasons, it is difficult to know how relevant the extracted wake deficit profiles would be for research purposes, or how well they would correlate with wake deficits found in previous research. Wake length was instead examined as a proxy to discern the downstream extent of the wake effect in both the convective and stable ABLs.

To more clearly convey differences in wind speed (and atmospheric turbulence) between the convective and stable ABLs, the convective and stable portions of the SD analysis period are now shaded in Fig. 15 (now Fig. 16). The revised figure was copied below for reference.
Pg. 26: ln. 14: ‘access to the controller design so any factors inhibiting proper implementation of the turbine control offsets can be identified.’ I agree that access to the controller improves the assessment of wind plant control strategies, and is always desirable, but I think meaningful control assessments can be done without direct access. For example, adding a pitch offset in region 2 (where pitch is typically fixed at ‘fine pitch’) could be achieved without needing to understand the controller dynamics. Furthermore, to implement a yaw misalignment, the yaw controller setpoint could be changed from zero to the desired offset, but full understanding of the controller dynamics is not necessary, and in many cases would be asking too much given the proprietary nature of wind turbine control systems.

Meaningful wind plant control experiments could be performed without full access to wind turbine controller design. However, a comprehensive analysis of the controller’s ability to implement the desired control strategies would likely require access to its control structure and design. Without this information, it would be difficult to discern why the turbine was able/unable to implement the desired control changes and it would also be challenging to determine how these control changes could be more efficiently employed. However, manuscript text was modified to soften reader interpretation.

Pg. 28 Lns. 20 through 21:
‘...(2) access to relevant controller information (e.g. controller design) so any factors inhibiting proper implementation of the turbine control offsets can be identified.’