Supplement of

Wind turbine power production and annual energy production depend on atmospheric stability and turbulence

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Supplement

S1 Lidar variability

Lidar measurements of wind speed and direction exhibit larger variability than those from the met tower. This variability may be due to the lidar’s operating assumption of homogeneity across the measurement volume: the WINDCUBE v1 measures volumetric-averaged wind speeds and directions over a 20 m thick layer with an effective diameter on the order of the height of the measurement (30° beam angle) and assumes homogeneous flow within that layer. This assumption may not be reliable at this site: observations from scanning lidar of flow at the U.S. Department of Energy (DOE) National Wind Technology Center (NWTC) at the National Renewable Energy Laboratory (NREL) indicate that flow can be very inhomogeneous (Smalikho et al., 2013; Aitken et al., 2014).

Despite this potential for variability at the NWTC, the lidar and tower measurements are generally well-correlated ($R = 0.96$ and bias of 0.37 m s$^{-1}$, with the lidar recording higher values than the anemometer for this time period; Fig. 8). Note that the bias is likely significantly influenced by the accuracies of the instrument compared (see Sect. 2.2.1 and Sect. 2.2.2). Previous work by Smith et al. (2006), Sathe et al. (2011) and Sanz Rodrigo et al. (2013) saw strong correlations between lidars and anemometers in flat terrain. Smith et al. (2006) found a correlation coefficient of 0.9843 between a ZephIR lidar and a cup anemometer at 80 m for 10-min wind speed averages for 1 day, Sathe et al. (2011) found correlation coefficients greater than 0.98 between WINDCUBE and ZephIR lidars and sonic anemometers at 100 m for 10-min wind speed averages for 4–5 months, and Sanz Rodrigo et al. (2013) found correlation coefficients greater than 0.99 between WINDCUBE and ZephIR lidars and cup anemometers at 89 m for 10-min wind speed averages for 10 days. Sanz Rodrigo et al. (2013) also performed lidar–tower comparisons in complex terrain in the Alaiz mountain range in Navarra, Spain, and found correlation coefficients greater than 0.98 between WINDCUBE and ZephIR lidars and cup anemometers at 78 m for 10-min wind speed averages for approximately 5 months. Our correlations between a lidar and an anemometer, based on 2.5 months of collecting wind speed and direction-filtered data in complex terrain in an atmosphere with relatively few aerosols for backscatter, resulted in a relatively high correlation coefficient of 0.96. Our correlation in an inhomogeneous flow is only slightly lower than other correlation coefficients previously found between lidars and towers in flat terrain.

S2 Rotor equivalent wind speed
Quantifying the wind profile across the entire swept rotor area (SRA) has been shown to improve correlations between wind inflow and power output (Wagner et al., 2009). Here, we calculate rotor equivalent wind speeds (REWS) following Wagner et al. (2009):

\[
U_{eq} = \left( \sum_i U_i^3 \frac{A_i}{A_{tot}} \right)^{1/3}, \tag{S1}
\]

where \(i\) represents the index of the level, \(U\) is the horizontal wind speed, \(A_i\) is the area of the turbine rotor disk of the level with the corresponding data point (the area of the sector defined by chord/arc relative to 360°, minus the area of the triangle), and \(A_{tot}\) is the SRA. When calculating the REWS from the lidar profiles, five levels (40, 60, 80, 100, 120 m) are available; when calculating the REWS from the tower cup anemometer data, three levels (55, 80, 105 m) are available, all of which use Thies anemometers. Cup anemometer data available at 38, 87, and 122 m are not used in these REWS calculations for consistency because these are different instruments than those at 55, 80, and 105 m, which were are to calculate the REWS.

Despite the variability in the shear exponent as calculated from the tower measurements (Sect. 3.7), the high correlations between the 80-m wind speeds and REWS shown in Fig. S1 suggest that power curves for 80 m will be very similar to power curves for REWS for this data set. The high turbulence at the NWTC may have prevented the occurrence of larger wind shear across the rotor disk. This might lead to significant differences in REWS from 80-m wind speeds, which may manifest in the power curves at other sites. Differences between the REWS and the wind speed at hub height at other sites may also affect annual energy production calculations as in Scheurich et al. (2016).

S3 Yaw error and veer

Additionally, we explore the effects of yaw error and wind veer and distributions of these variables as shown in Fig. S2. To calculate yaw error, we subtract the wind direction as measured by the met tower near hub-height from the nacelle position as given by the supervisory control and data acquisition (SCADA) system. The resulting yaw error, however, is centered around 90° instead of 0°, which means the orientation of the turbine position is around 90° off of North. To correct for this, we assume that the yaw error should be 0° at rated power, so we take the average yaw error when the turbine was producing rated power (94.22°) and subtract this from the yaw error to get the correct values of yaw error. After correcting for the turbine yaw orientation offset, we determined that it is not appropriate
to split the yaw error distribution into regimes as 78% of the data lie within ± 5° yaw error and 96% of the data lie within ± 10° yaw error.

We found no impact of yaw error or wind veer on the power curves at this site. However, based on other locations (Vanderwende and Lundquist, 2012; Rhodes and Lundquist, 2013; Walton et al., 2014) where significant veer does occur, it may affect power production, but this site does not regularly experience that phenomena.

**S4 Power curves for different TKE regimes**

Likely due to a lack of sonic data at 74 m that passes our data quality control filters as discussed in Sect. 3.1 (60 %), few statistically-distinct bins emerge from the TKE power curves. Fig. S3 shows the nacelle and upwind tower power curves segregated by TKE regime. Only at about 12 m s⁻¹ do statistically-significant differences in power curves emerge between the low and high TKE power curves: at 12 m s⁻¹, cases within the low TKE regime produce significantly more power than cases within the high TKE regime.

**References**


Supplementary Figures

**Figure S1.** REWS as a function of 80-m wind speed from (a) the tower and from (b) lidar. Black dotted line represents a 1:1 relationship. Includes data filtered for tower 80-m wind speeds between 3.5 and 25.0 m s\(^{-1}\), 87-m wind directions between 235° and 315°, and for normal turbine operation.
Figure S2. (a) Yaw error histogram and (b) wind veer histogram. Includes data filtered for tower 80-m wind speeds between 3.5 and 25.0 m s$^{-1}$, 87-m wind directions between 235° and 315°, and for normal turbine operation.
Figure S3. (a) Nacelle anemometer and (b) 80-m tower anemometer power curves with TKE regimes. (c) Nacelle anemometer and (d) 80-m tower anemometer power curves shown as the anomaly from the neutral or medium power curve of the TI regimes. Median statistics are used to avoid outlier effects. Statistically distinct differences within each wind speed bin between the regimes are determined by the Wilcoxon rank sum test with a 1% significance level and denoted by closed circles. Includes data filtered for tower 80-m wind speeds between 3.5 and 25.0 m s⁻¹, 87-m wind directions between 235° and 315°, and for normal turbine operation. Envelopes represent ±1 MAD for each wind speed bin. The grey dashed line marks rated speed.