

Wind Vane Correction during Yaw Misalignment for Horizontal Axis Wind Turbines

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Abstract. This paper investigates the accuracy of wind direction measurements for horizontal axis wind turbines and [\[..¹ \]their](#) impact on yaw control. The yaw controller is crucial for aligning the rotor with the wind direction and [\[..² \]optimising](#) energy extraction. Wind direction is conventionally measured by one or two wind vanes located on the nacelle, but the proximity of the rotor can interfere with these measurements. The authors show that the conventional corrections, including low-pass filters and calibrated offset correction, are [\[..³ \]inadequate to correct](#) a systematic overestimation of the wind direction deviation caused by the rotor misalignment. This measurement error can lead to an overcorrection of the yaw controller [\[..⁴ \]and](#), thus, to an oscillating yaw behaviour, even if the wind direction is relatively steady. The authors present a theoretical basis and methods for quantifying the wind vane measurement error and validate their findings using computational fluid dynamics simulations and operational data from two commercial wind turbines. Additionally, the authors propose a correction function that improves the wind vane measurements and demonstrate its effectiveness in two [\[..⁵ \]free-field](#) experiments. Overall, the paper provides new insights into the accuracy of wind direction measurements and proposes solutions to improve the yaw control for horizontal axis wind turbines.

1 Introduction

Wind turbines are [becoming](#) an increasingly important [\[..⁶ \]renewable energy source](#), and their [widespread adoption depends on their](#) performance and efficiency[\[..⁷ \]](#). [One crucial factor affecting the performance of](#) horizontal axis wind [\[..⁸ \]turbines](#) is the alignment of the rotor with the incoming wind direction. The [\[..⁹ \]turbine can only achieve its highest possible power coefficient when facing directly into the wind](#). Even minor misalignment can impair its ability to convert the

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wind's kinetic energy into electricity and result in uneven forces on the blades, leading to increased fatigue loads. In certain circumstances, the turbine's intentional misalignment can manipulate the wake to reduce the impact on downstream turbines. This approach is called Active Wake Deflection, and we will discuss it briefly below. However, even for this specific mode of operation, accurately estimating the wind turbine's alignment is crucial. The standard procedure to determine the alignment involves using one or two wind vanes to detect deviations from the wind direction and adjust the yaw angle of the turbine through an active yaw manoeuvre accordingly. ^[..¹⁰]During commissioning, the wind vanes are oriented along the rotor axis, followed by offset correction calibration to account for wake rotation over the nacelle, thereby achieving the most precise alignment of the wind turbine in the flow direction for a wind direction deviation of 0°. The IEC 61400-12-3 standard for ^[..¹¹]measurement-based site calibration (International Electrotechnical Commission, 2022) provides guidelines for ^[..¹²]measuring, analysing, and reporting ^[..¹³]site calibration in power performance testing for wind turbines. ^[..¹⁴]However, due to the intricate flow field surrounding the turbine and associated wake effects, this still poses a significant challenge. The question arises as to whether it is possible to reduce uncertainties and calibrate the wind vane ^[..¹⁵]to achieve better wind turbine alignment without resorting to additional or external measuring systems. In this way, on the one hand, a higher power yield could be achieved, and, on the other hand, the yaw activity could be reduced, which would protect the yaw motors and brakes and, thus, increase their lifetime.

In (Kragh and Fleming, 2012), an amplification of the wind direction deviation behind the rotor of a test turbine was pointed out, which is correlated to the rotor speed. A linear correction function depending on the rotor speed was presented, which was used to improve the measured wind direction deviation of the test turbine. Mittelmeier and Kühn (2018) presented a ^[..¹⁶]three-step method based on SCADA data to detect changes in the wind turbine alignment during the operational lifetime and to improve the alignment.

Additional temporary or permanent measurement devices, such as spinner anemometry by ^[..¹⁷]ultrasonic sensors (Pedersen et al., 2014), nacelle-based lidars (Held and Mann, 2019) or monitoring of cyclic blade root bending moments (Bertelè et al., 2017; Schreiber et al., 2020) have been proposed to improve the alignment.

A subject of extensive scientific research for several years has been the active wake deflection ^[..¹⁸](Gebraad et al., 2016; Rott et al., 2018; Bromm et al., 2018). Especially for this kind of control, a well-calibrated wind vane is essential, as this method, in particular, requires specific misalignments to be maintained.

Recently Simley et al. (2021) has reported that during experiments on a Senvion MM82 wind turbine, it was observed that the wind vane overestimated the wind direction deviation compared to ^[..¹⁹]nacelle-based lidar measurements. A linear wind-

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speed-dependent transfer function was proposed to correct the wind vane. Simley et al. suggested that more complex functions may be more appropriate for future yaw controllers, and more research is needed in this area.

Nevertheless, many questions remain unanswered, such as the occurrence and causes of deterministic errors of the wind vane and how the wind vane could be calibrated or corrected to achieve better performance for regular operation or specific control techniques such as active wake deflection.

^[..20] Our study investigates the yaw behaviour of two commercial wind turbine types. We observe the wind vane signal before and after a yaw manoeuvre and compare the obtained wind direction with a reference signal from a nearby measuring mast. In addition, we conduct a multi-stage experiment with the wind turbine, in which we investigate different correction functions for the wind vane. With this publication, we would like to find answers to the following questions:

1. ^[..21] Is there a systematic error in wind vane readings when utility-scale wind turbines ^[..22] are not aligned with the wind direction, and how can ^[..23] this error be described?
2. ^[..24] Can wind vane measurements be corrected using operational data, both with and without external reference measurements?
3. ^[..25] How does correcting the wind vane during yaw misalignment ^[..26] affect a wind turbine's performance?

2 Methods

In this section, we first give a brief overview of the operation of a conventional wind turbine yaw controller (Section 2.1). Then, we outline our hypothesis about the causes leading to a measurement error of a wind vane behind the rotor of a wind turbine and create a model for the error estimation (Section 2.2). Next, we detail the Computational Fluid Dynamics (CFD) simulation we performed to confirm our assumption (Section 2.3). In subsection (Section 2.4), we ^[..27] describe the free-field data we used and ^[..28] our experiments with two commercial wind turbines.

2.1 General Yaw Control

^[..29] Active yaw control is commonly applied to ensure a wind turbine aligns with the wind direction^[..30]. As described in the Wind Energy Handbook (Burton et al., 2011), from the yaw error measured by the wind vane on the nacelle^[..31], a demand

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signal for the yaw actuator is calculated. To avoid the yaw control being influenced by small fluctuations, the measurement of the wind vane is averaged (e.g. moving averages with a window size of 30s, 60s or even 180s are commonly used). [..³²] A dead-band controller is typically used for the control, where a yaw manoeuvre is initiated if the yaw error exceeds a predefined threshold. We refer to this threshold as the *yaw trigger*. With a standard yaw control, the magnitude of the yaw rotation corresponds to the determined yaw error. However, [..³³] this value may differ from the measured deviation for particular yaw strategies. An example [..³⁴] is the active wake deflection [..³⁵] (Gebraad et al., 2016; Rott et al., 2018) mentioned above, in which a specific yaw misalignment is applied for certain wind direction sectors in a wind farm. Since the target value of the yaw manoeuvre is also adjusted in this study, we refer to this value as the *yaw target*.

2.2 Wake Deflection

As a motivation for the investigation carried out in this study, we look at an example time series of the wind direction measurement by a wind vane on the nacelle of a commercial wind turbine. The wind vane measures the wind direction deviation from the nacelle's orientation, which we denote by $\varphi_{wt}(t) \in [-180^\circ, 180^\circ)$, where $t \in \mathbb{R}$ represents the time. The measured wind direction in the global frame of reference $\omega_{wt}(t) \in [0^\circ, 360^\circ)$ is the sum of the wind direction deviation $\varphi_{wt}(t)$ and the orientation of the nacelle (yaw angle) $\gamma_{wt}(t) \in [0^\circ, 360^\circ)$:

$$\omega_{wt}(t) \equiv \varphi_{wt}(t) + \gamma_{wt}(t) \pmod{360^\circ}. \quad (1)$$

For the sake of better readability, we will omit the modulo notation in the following and imply that angle specifications always lie in the value range $[-180^\circ, 180^\circ)$ for angular deviations and in the value range $[0^\circ, 360^\circ)$ for absolute angles. Figure 1 shows $\gamma_{wt}(t)$, $\omega_{wt}(t)$, and the wind direction with a centred 60-s moving average:

$$\tilde{\omega}_{wt}(t) = \frac{1}{60\text{s}} \int_{t-30\text{s}}^{t+30\text{s}} \omega_{wt}(\tau) d\tau. \quad (2)$$

[..³⁶] We have used the arithmetic mean for the directional values here and [..³⁷] in the following [..³⁸] rather than the directional mean calculated over the vectorial components, even though this is not technically correct. We have chosen to use the arithmetic mean because this is more widely used for such calculations when dealing with Supervisory Control and Data Acquisition (SCADA) data [..³⁹] and because the difference is negligible for the relatively small directional values around 0° , as is the case here.

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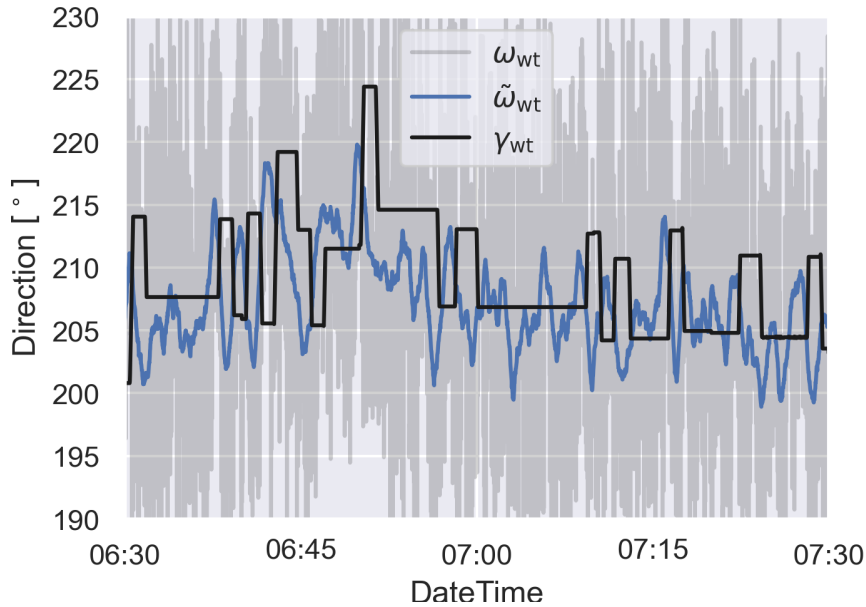


Figure 1. Example of a 60-min time series from a BARD 5.0 wind turbine of the nacelle orientation γ_{wt} (black) and the wind direction measured by the wind vane ω_{wt} in 1-Hz resolution (grey). In addition, the wind direction was averaged with a [\[.40\]](#) centred 60-s moving window $\bar{\omega}_{wt}$ (blue).

In Figure 1, it is noticeable that the BARD 5.0 wind turbine makes a relatively large number of yaw manoeuvres in the 60-min time series, even though the measured averaged wind direction $\bar{\omega}_{wt}(t)$ changes relatively little in the [\[.41\]](#) period shown. In most cases, the directions of the yaw manoeuvre alternate, i.e. a clockwise rotation (yaw angle increases) is followed by a counterclockwise rotation of the nacelle (yaw angle decreases). Furthermore, although the wind direction looks like a highly noisy stochastic process, i.e., it undergoes random and unpredictable changes, it seems that the moving average of the wind direction [\[.42\]](#) turns in the opposite direction [\[.43\]](#) precisely in situations where the wind turbine performs a yaw manoeuvre. In Layman’s terms, it is almost as if the wind is trying to avoid the wind turbine. It should be noted, however, that the time series shown here has been deliberately chosen so that these features can be easily identified. Nevertheless, we could observe this behaviour [\[.44\]](#) often when examining measurement data from different wind turbines [\[.45\]](#). Therefore, we presume a causality exists between the measured wind direction changes and the turbine, which [\[.46\]](#) motivated the present investigation.

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[..⁴⁷] We hypothesise that the exact mechanism that causes the deflection of the intermediate to far wake during yaw misalignment (Jiménez et al., 2010; Bastankhah and Porté-Agel, 2016), i.e. the thrust component perpendicular to the inflow direction, is [..⁴⁸] also affecting the wind direction measured a few metres behind the rotor plan on top of the nacelle. Such a deflection, though, would explain an overestimation of the wind direction deviation by a wind vane on the nacelle. Due to an overestimation of the wind direction deviation, a yaw manoeuvre is triggered in the yaw algorithm earlier than intended and the orientation to which the rotor adjusts overshoots the actual target. This increases the probability that the wind turbine [..⁴⁹] has an opposite yaw [..⁵⁰] misalignment. If the wind turbine again overestimates this resulting misalignment, this can result in an alternating yaw pattern, where the wind turbine tries to follow the wind direction [..⁵¹] but repeatedly overshoots.

In order to model the overestimation of the wind vane, we use a simple linear transfer function [..⁵²] that approximates the relationship between the wind direction measured by the wind vane $\varphi_{wt} \in [-180^\circ, 180^\circ]$ and an estimate (represented by the hat $\hat{\cdot}$) of the "true" or reference wind direction deviation $\varphi_{ref} \in [-180^\circ, 180^\circ]$:

$$\hat{\varphi}_{ref} = c \cdot \varphi_{wt} + b, \quad (3)$$

where $c, b \in \mathbb{R}$ are the parameters describing the slope and the offset, respectively. The offset b is attributed to [..⁵³] the wind vane [..⁵⁴]’s mounting error and the wake’s rotation. It is usually determined during the calibration of the wind vane or by more elaborate analyses of the power performance (Mittelmeier and Kühn, 2018). [..⁵⁵] As a result, wind vane data typically incorporates an adjustment to account for this deviation. Consequently, our subsequent analysis focuses on the correction factor c [..⁵⁶], with the offset factor [..⁵⁷] b being set to 0° .

In the following sections, we will show how we estimate the correction factor c [..⁵⁸] by comparing the turbine’s wind vane and a met mast and [..⁵⁹] only the SCADA signal measurements at the turbine.

2.3 CFD Simulation Setup

For [..⁶⁰] verification, CFD simulations of the NREL 5MW reference turbine were performed (Jonkman et al., 2009) using the open-source CFD software OpenFOAM (OpenFOAM, 2021). The numerical grid was generated [..⁶¹] using the two in-house

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125 tools *bladeBlockMesher* and *windTurbineMesher* (Rahimi et al., 2016)[⁶²], and consists of 26.4 mio cells. The rectangular meshes contain the yawed rotor with a diameter of $D = 126\text{m}$ and a cylindrical, non-rotating nacelle geometry, neglecting the influence of the tower. The length of the nacelle was chosen to be 16 m with a radius of 1.35 m. The rotor is located $5D$ from the inlet and $15D$ from the outlet, with a distance of $3.5D$ towards all sides. Extra mesh refinement was made in the vicinity of the nacelle and blade roots[⁶³], ensuring that the flow at the probe locations is resolved reasonably well. The rotation of the rigid blades is accounted for using sliding mesh interfaces between the rotor and the [⁶⁴]far-field grids. Five [⁶⁵]yaw angles were investigated, namely, $-20^\circ, -10^\circ, 0^\circ, 10^\circ$ and 20° , with an inflow wind velocity of 11.4m/s and a constant rotational speed of 12.1 RPM.

The incompressible, transient flow was simulated using the hybrid Spalart-Allmaras delayed detached eddy simulation type (Spalart et al., 2006). [⁶⁶]A second-order implicit backward method was used to advance the solution in time. Temporal discretization made use of a second-order accurate Gauss linear scheme.

On top of the nacelle, a total of 390 probes were placed at three different heights (2.35 m, 2.93 m and 3.50 m) above the axis of rotation, representing possible wind vane positions. Figure 2 provides a schematic view of the turbine and probe positions from both the top and the side [⁶⁷]and Figure 3 displays a perspective view from the simulation. In both figures, the axis origin is shifted from the [⁶⁸]centre of the rotor in front of the turbine to make it visible.

140 2.4 Free Field Data

Our second approach to verify our hypothesis and find the parameters for our linear model (Eq. (3)) is to examine measurements from commercial turbines and [⁶⁹]conduct experiments in the free field. For this, we used two sets of data. Firstly, measurement data of the BARD 5.0 wind turbine, [⁷⁰]located at the prototype site in [⁷¹]northwest Germany on the Rysumer Nacken consisting of two wind turbines of this type. The turbine has a rotor diameter of 122m, a hub height of 90m and a rated power of 5MW. More details on the BARD 5.0 wind turbine can be found in Teubler (2011). [⁷²]Secondly, data from the eno114 wind turbine from the Kirch Mulsow test field in north Germany. For more information regarding the eno114 and the test field Kirch Mulsow, see (Hulsman et al., 2022).

At both locations, the measurements of the wind vane could be compared with a mast [⁷³]set up at a distance of approx. 300 m in each case. At the eno114, however, we were able to analyse situations within the scope of an investigation into wake deflection, in which the rotor was intentionally misaligned to the wind direction by up to 20° .

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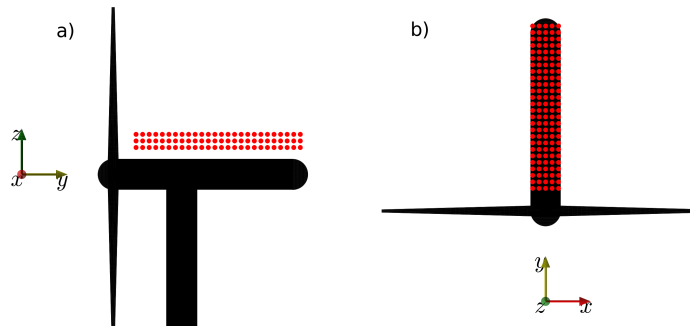


Figure 2. Sketch of the setup of the 390 probe positions (in red) from the side view a) and the top view b). The axis origin is located directly in the rotor centre and has been moved in front of the rotor for better visibility.

2.4.1 Comparison between Wind Vane and Met Mast

[..⁷⁴] To identify an error or an overestimation of the wind vane, we compare the [..⁷⁵] turbine's wind vane **measurements** with the wind direction measurements of the meteorological mast (met mast) as a reference.

At the Rysumer Nacken test site, there are two wind vanes installed on the BARD5.0 wind turbine at [..⁷⁶] **the** height of 155 91 m approximately 8 m behind the rotor plane at a lateral distance of 2 m for the rotor axis and 1.5 m above the nacelle. One of the type *Wind direction Sensor INDUSTRY* (00.14567.110040) from Lambrecht meteo GmbH and one of the type *Ultrasonic Anemometer 2D compact* (4.387x.xx.xxx) by Thies Clima. The [..⁷⁷] **two-system** setup serves to increase availability. When both devices are operational, the yaw control uses the mean value of both wind vanes, which is also the measurement we are investigating. The met mast is equipped with a wind vane at a height of 90 m.

160 At the wind farm in Kirch Mulsow, one ultrasonic wind vane of the type *Ultrasonic Anemometer 2D compact* (4.387x.xx.xxx) by Thies Clima is installed on the eno114 wind turbine in 120 m height approximately 12 m behind the rotor and 1.5 m above

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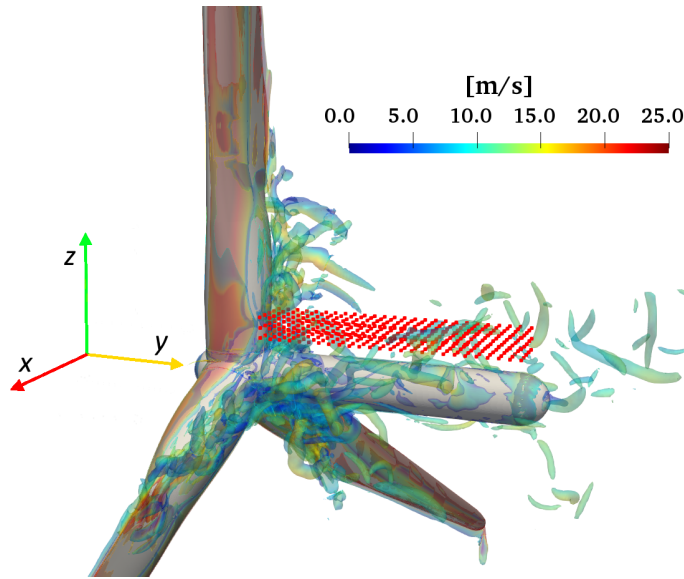


Figure 3. Nacelle region of the simulated rotor with the 390 probe locations marked in red. The instantaneous root vortices that influence the probe region are described using an iso-surface of the λ_2 criterion of Jeong and Hussain (1995) for a value of $\lambda_2 = 100 \frac{1}{s^2}$ and are coloured by the velocity magnitude.

the nacelle. The met mast uses a wind vane of the type Wind Direction Transmitter "First Class" (4.3151.00.x1x) from Thies Clima at a height of 112 m.

165 Analogous to the above, we refer to the wind direction deviation measured by the turbine as $\varphi_{wt}(t) \in [-180^\circ, 180^\circ]$. To compensate for [..⁷⁸] small-scale fluctuations, we resample the measurements to 60-s averages. We denote the wind direction measured by the met mast with $\omega_{mm}(t) \in [0^\circ, 360^\circ]$, from this we calculate the wind direction deviation of the wind turbine determined by the met mast as $\varphi_{mm}(t) := \omega_{mm}(t) - \gamma_{wt}(t)$ and also resample the data to 60-s averages.

170 [..⁷⁹] To determine the influence of the [..⁸⁰] rotor's thrust on the wind vane [..⁸¹] of the wind turbine, only situations in which the wind turbine [..⁸²] was operated in the partial load range and without curtailment were [..⁸³] considered to compare the measured values.

[..⁸⁴] We use an "Orthogonal Distance Regression" (ODR) (Boggs and Rogers, 1990) to quantify the relationship between the two measured values. This regression method works similarly to an "Ordinary Linear Regression" (OLS). In both methods, the measured values are transferred as tuples into a coordinate system (scatter plot), in our case $(\varphi_{wt}(t), \varphi_{mm}(t))_t$. In an OLS,

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an affine linear function is determined that minimises the squared distance of the function values to the dependent variable of the data points, which is the second component of the ^[..⁸⁵]tuple and which is usually plotted on the vertical axis (y -axis) of a graph. This is reasonable if the first component of the ^[..⁸⁶]tuple, which is ^[..⁸⁷]generally called the predictor variable or independent variable and is usually plotted on the horizontal axis (x -axis) of a graph, does not ^[..⁸⁸]depend on other factors and has no uncertainty. One of the most common examples ^[..⁸⁹]is if the independent variable is the timestamp. However, if the predictor has uncertainties, the ^[..⁹⁰]regression slope will be biased towards zero. This phenomenon is ^[..⁹¹]referred to as "regression dilution" and is extensively discussed in (Frost and Thompson, 2000). Another ^[..⁹²]drawback of the OLS (Ordinary Least Squares) technique is its lack of invertibility. This implies that if you interchange the x and y data and implement OLS on the transformed tuples, the ^[..⁹³]resultant linear function will not be the inverse of the ^[..⁹⁴]initial linear function. In contrast, an ODR minimises the squared orthogonal distances from a regression function to the tuples. The regression line obtained by this method accounts for uncertainties ^[..⁹⁵]in the data points' first and second components. With an ODR, the variables can be swapped, thereby inverting the regression line. It should be noted, however, that the uncertainties in both measured values are weighted equally. It is possible to obtain a different weighting for the uncertainties of both measured values by stretching or compressing the data on one of the axes of the coordinate system. ^[..⁹⁶]However, we assume that the 60-s mean values considered have similar uncertainties. The gradient of the ODR regression line gives us an estimate for the correction factor c . The results of this investigation are presented in section 3.2.1.

190 2.4.2 Yaw Manoeuvre Analysis

^[..⁹⁷]This section introduces a method for estimating the ^[..⁹⁸]wind vane's correction factor without needing external measurements, such as a measuring mast. Similar to the step response analysis for time-invariant linear systems, we compare the wind vane measurements immediately before a yaw manoeuvre with those after.

First, we filter the SCADA data to exclude situations ^[..⁹⁹]where the wind turbine is not generating electricity or ^[..¹⁰⁰]operating at reduced output. Then, we identify all yaw manoeuvres in the SCADA data and divide them into clockwise

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(cw) and counterclockwise (ccw) yaw manoeuvres. On the one hand, the distinction between the two directions of rotation is [\[.101\]](#) essential for data processing since the different signs for both directions of rotation must be [\[.102\]](#) considered in the statistical evaluations. [\[.103\]](#) Furthermore, a correction factor for both directions of rotation can be determined [\[.104\]](#) through the differentiation. Thus, a possible asymmetry can be detected, which [\[.105\]](#) may be caused, e.g. by an offset error, which was not sufficiently eliminated [\[.106\]](#) during the calibration of the wind vane. Regular yaw manoeuvres usually take less than 30 seconds. In some cases, however, a yaw manoeuvre can take longer [\[.107\]](#). This indicates that the wind turbine is realigning itself after a shutdown, that a cable de-twist is taking place or that the wind direction is abruptly changing very strongly. [\[.108\]](#) We only consider yaw manoeuvres that lasted less than 30 seconds to exclude these cases from our investigations. In the following, we restrict ourselves to describing the methods for the cw yaw manoeuvres [\[.109\]](#) since the techniques are used analogously for evaluating the ccw yaw manoeuvres.

The time at which the i -th cw yaw manoeuvre ($i \in \mathbb{N}$) starts is denoted as $t_{ys,i} \in \mathbb{R}$ and the time at which it ends $t_{ye,i} \in \mathbb{R}$ (ys and ye denote "yaw start" and "yaw end", respectively). The number of all cw yaw manoeuvres obtained in this way is denoted as n_{cw} . For [\[.110\]](#) empirical data analysis, we consider the measurements over a period of the [\[.111\]](#) size $T \in \mathbb{R}$ before [\[.112\]](#) the start of the cw yaw manoeuvre $([t_{ys,i} - T, t_{ys,i}])_i$ and respectively after the end of the yaw manoeuvre $([t_{ye,i}, t_{ye,i} + T])_i$. The length of the time interval T must be selected sufficiently small, depending on the configuration of the yaw controller, so that no further yaw manoeuvres within the time interval interfere with the measurements and the yaw angle is constant before and after the yaw manoeuvre, respectively. T must be selected large enough to suppress turbulence-related measurement noise as far as possible. For our investigations, we therefore chose $T = 60$ s. The measurements during the yaw manoeuvres $[t_{ys,i}, t_{ye,i}]$ are not considered in the analysis because the duration of the yaw manoeuvres varies, and there are additional uncertainties during the [\[.113\]](#) nacelle rotation. For the selected periods, we now consider the wind direction measurements of the wind turbines. For aggregating the data, we [\[.114\]](#) centre the measurements around the yaw angle at the end of the respective yaw manoeuvre and thus obtain the following expression:

$$\bar{\omega}_{wt,cw}(\tau) = \begin{cases} \frac{1}{n_{cw}} \sum_{i=1}^{n_{cw}} \omega_{wt}(t_{ys,i} + \tau) - \gamma_{wt}(t_{ye,i}), & \text{for } \tau \in [-T, 0\text{s}] \\ \frac{1}{n_{cw}} \sum_{i=1}^{n_{cw}} \omega_{wt}(t_{ye,i} + \tau) - \gamma_{wt}(t_{ye,i}), & \text{for } \tau \in (0\text{s}, T] \end{cases}. \quad (4)$$

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220 **[.115]Similarly**, we also average the yaw angles **[.116]centred** around the yaw angle at the end of the respective yaw manoeuvre:

$$\bar{\gamma}_{\text{wt,cw}}(\tau) = \begin{cases} \frac{1}{n_{\text{cw}}} \sum_{i=1}^{n_{\text{cw}}} \gamma(t_{\text{ys},i} + \tau) - \gamma(t_{\text{ye},i}) & \text{for } \tau \in [-T, 0\text{s}] \\ 0^\circ & \text{for } \tau \in (0\text{s}, T] \end{cases} \quad (5)$$

As already mentioned, the yaw angle is constant before the yaw manoeuvre $\bar{\gamma}_{\text{wt,cw}}(\tau) = \gamma_{\text{wt,cw}}^{\tau \leq 0}$ for $\tau \leq 0\text{s}$, only transitions during the yaw manoeuvre and is constant again after the yaw manoeuvre $\bar{\gamma}_{\text{wt,cw}}(\tau) = \gamma_{\text{wt,cw}}^{\tau > 0} = 0^\circ$ for $\tau > 0\text{s}$. The wind direction deviations are aggregated without **[.117]centring** according to:

$$225 \quad \bar{\varphi}_{\text{cw}}(\tau) := \begin{cases} \frac{1}{n_{\text{cw}}} \sum_{i=1}^{n_{\text{cw}}} \varphi_{\text{wt}}(t_{\text{ys},i} + \tau), & \text{for } \tau \in [-T, 0\text{s}] \\ \frac{1}{n_{\text{cw}}} \sum_{i=1}^{n_{\text{cw}}} \varphi_{\text{wt}}(t_{\text{ye},i} + \tau), & \text{for } \tau \in (0\text{s}, T] \end{cases} \quad (6)$$

From this, we can calculate the time-averaged deviation of the mean wind direction $\bar{\varphi}_{\text{cw}}$ before the yaw manoeuvre:

$$\bar{\bar{\varphi}}_{\text{cw}}^{\tau \leq 0} := \frac{1}{T} \int_{-T}^{0\text{s}} \bar{\varphi}_{\text{cw}}(\tau) d\tau, \quad (7)$$

and after the yaw manoeuvre:

$$\bar{\bar{\varphi}}_{\text{cw}}^{\tau > 0} := \frac{1}{T} \int_{0\text{s}}^T \bar{\varphi}_{\text{cw}}(\tau) d\tau. \quad (8)$$

230 Figure 4 gives a schematic illustration of these aggregated measurements. The wind direction $\bar{\omega}_{\text{wt,cw}}(\tau)$ is **[.119]shown** in light red together with the average yaw angle $\bar{\gamma}_{\text{wt}}(\tau)$ in blue. The bold red line represents the time-averaged measured wind direction before the yaw manoeuvre **[.120]**. **Therefore**, the difference between the bold red line and the yaw angle is $\bar{\bar{\varphi}}_{\text{cw}}^{\tau \leq 0}$, as shown in the figure. Analogously, the bold magenta line represents the time-averaged measured wind direction after the yaw manoeuvre, which means the difference between this value and the yaw angle is $\bar{\bar{\varphi}}_{\text{cw}}^{\tau < 0}$. **[.121]The centred "true" reference**
 235 wind direction $\bar{\omega}_{\text{ref}}(\tau)$ is shown **in green**. The reference wind direction cannot be measured directly and is **[.122]unknown**, but we can state the following two assumptions:

First, the measured and the true wind direction deviation can be expressed by the linear function from Eq. (3). The correction c is denoted as c_{cw} , since, in this case, we only analyse clock-wise yaw manoeuvres. **[.123]The figure shows this relation by**

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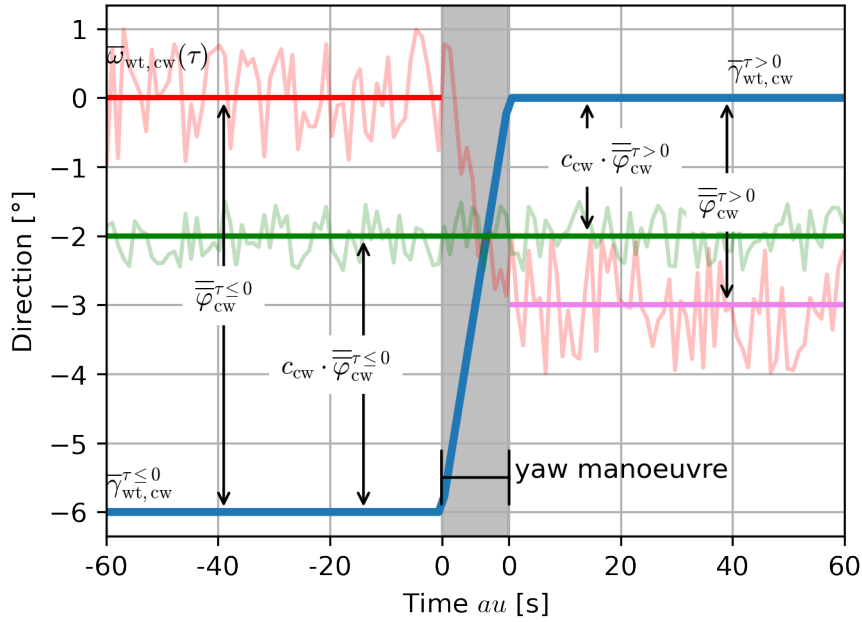


Figure 4. Schematic illustration of the average measured wind direction $\bar{\omega}_{wt,cw}(\tau)$ (in light red) before and after a yaw manoeuvre, the average yaw angle $\bar{\gamma}_{wt,cw}(\tau)$ (blue) and the average true wind direction $\omega_{ref}(\tau)$ (green) all [\[..118\]](#) centred at the respective yaw angle after the yaw manoeuvre.

the difference between the green line and the yaw angle before and the green line and the yaw angle after the yaw manoeuvre,
 240 respectively. Second, the wind direction is a stationary random process for the duration $\tau \in [t_{ys,i} - T, t_{ye} + T]$. Now we can postulate the following relationship:

$$\hat{\omega}_{ref,cw} = \bar{\gamma}_{wt,cw}^{\tau \leq 0} + c_{cw} \cdot \bar{\varphi}_{cw}^{\tau \leq 0} = \bar{\gamma}_{wt,cw}^{\tau > 0} + c_{cw} \cdot \bar{\varphi}_{cw}^{\tau > 0} . \quad (9)$$

Solving for c_{cw} gives us:

$$c_{cw} = \frac{\bar{\gamma}_{wt,cw}^{\tau > 0} - \bar{\gamma}_{wt,cw}^{\tau \leq 0}}{\bar{\varphi}_{cw}^{\tau \leq 0} - \bar{\varphi}_{cw}^{\tau > 0}} . \quad (10)$$

245 The yaw control ensures that the yaw angle after the yaw manoeuvre $\bar{\gamma}_{wt,cw}^{\tau > 0}$ corresponds to the measured wind direction before the yaw $\bar{\omega}_{wt,cw}^{\tau \leq 0}$ (see Eq. (1)). This holds [\[..124\]](#) especially for the aggregated averages considered here, therefore: $\bar{\gamma}_{wt,cw}^{\tau > 0} = \bar{\omega}_{wt,cw}^{\tau \leq 0} = \bar{\gamma}_{wt,cw}^{\tau \leq 0} + \bar{\varphi}_{cw}^{\tau \leq 0}$. The correction factor can thus be estimated only by the wind vane measurements:

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$$c_{cw} := \frac{\overline{\varphi}_{cw}^{\tau \leq 0}}{\overline{\varphi}_{cw}^{\tau \leq 0} - \overline{\varphi}_{cw}^{\tau > 0}}. \quad (11)$$

We proceed analogously for ccw yaw manoeuvres.

250 3 Results

[..¹²⁵] This section presents the results of the methods described in the previous section [..¹²⁶]. Since multiple different datasets were analyzed for these studies, there is a short overview here:

- In Section 3.1, the data from the CFD simulation is analyzed as described in Section 2.3.
- In [..¹²⁷] Section 3.2, the methods described in Section 2.4 are applied to measured data from the free field.
- 255 – In Subsection 3.2.1 the wind wane measurements from the BARD5.0 at the Rysumer Nacken and the eno114 at Kirch Mulsow are compared to wind direction measurements [..¹²⁸] from the met mast at their respective locations as described in Section 2.4.1 .
- In Subsection 3.2.2 the yaw step analysis described in Section 2.4.2 is applied to the BARD5.0 wind turbine at Rysumer Nacken.
- 260 – And finally 3.3.1 [..¹²⁹] show results from a free field experiment where the correction factor was applied to the yaw controller of the BARD5.0 wind turbine at Rysumer Nacken.

3.1 CFD Simulation

From the data generated by the CFD simulation, we omitted the first 10 seconds of the simulation in our evaluation, in which the wake develops directly behind the rotor. In addition, we only consider measuring points at a minimum distance of 5 m
 265 behind the rotor plane, [..¹³⁰] meaning 300 of the 390 probes were used for the evaluation. The measured values were averaged over the three different measurement heights since [..¹³¹] observing the individual measurement heights did not reveal any special features. Figures 5 a) to e) show the time-averaged wind directions for the different measurement points behind the rotor in colour.

The plots display the deviations of the wind directions from the incoming wind direction coming from the south. A red
 270 colour indicates a deflection to the right (positive [..¹³²]), and a blue colour to the left ([..¹³³] negative). In Figure 5 a) the wind

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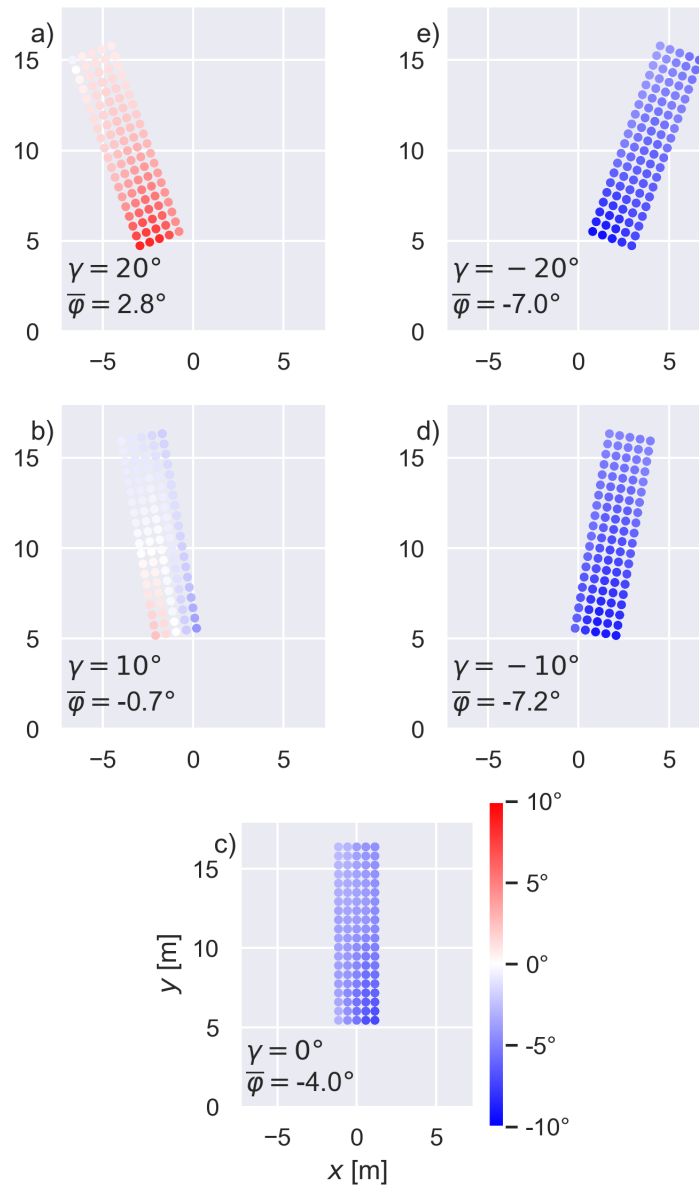


Figure 5. Coloured representation of the averaged wind direction above the nacelle for five different misalignments $\gamma \in \{-20^\circ, -10^\circ, 0^\circ, 10^\circ, 20^\circ\}$. A red colouring represents a deflection to the right (positive), and a blue colouring a deflection to the left (negative). The mean wind direction of all measurement points over the whole simulation time is given as $\bar{\varphi}$.

turbine is yawed by 20° (counterclockwise) relative to the inflow. The flow at the measurement locations is [¹³⁴]relatively

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homogeneous, with a ¹³⁵more substantial deviation closer to the ¹³⁶rotor plane. The average flow at the measurement locations has a direction of $\bar{\varphi} = 2.8^\circ$. ¹³⁷Therefore, the total average misalignment a virtual wind vane would experience is ¹³⁸22.8°. Figure 5 b) depicts the ¹³⁹wind turbine yawed by 10° (counterclockwise) relative to the inflow. The flow at the measurement locations shows ¹⁴⁰minimal deviation from the inflow. A ¹⁴¹slight positive trend on the ¹⁴²right-hand side of the figure close to the rotor plane and a negative trend on the ¹⁴³left-hand side of the figure can be observed. The overall ¹⁴⁴flow average at the measurement locations is $\bar{\varphi} = -0.7^\circ$, so the total misalignment the virtual wind vane would measure is 9.3° . In Figure 5 c) the wind turbine is aligned with the inflow. ¹⁴⁵A deviation to the left with an average of $\bar{\varphi} = -4.0^\circ$ can be seen at the measurement locations. This shows the ¹⁴⁶deviation mentioned above due to the rotation of the blades, which ¹⁴⁷causes a counter-rotation of the flow behind the rotor plane. In Figures 5 d), where the wind turbine is misaligned by -10° (i.e. 10° in a clockwise direction), this deflection gets even ¹⁴⁸more potent with an average flow direction of $\bar{\varphi} = -7.2^\circ$. A virtual wind vane would experience ¹⁴⁹a misalignment of -17.2° on average. Similarly in the last Figure 5 e) the misalignment is comparable to the case before, with an average misalignment of $\bar{\varphi} = -7.0^\circ$. A virtual wind vane would experience a misalignment of -27.0° .

Figure 6 summarizes the average virtual measured misalignments (blue dots), additionally ¹⁵⁰an ordinary least square (OLS) regression was fitted to the measurements (red line), ¹⁵¹which shows a very strong linear trend with a correlation coefficient of $R = 0.998$. Here, it can be observed that at the measuring points, an amplification of the actual misalignment is measured in the magnitude of 26% (the slope is $m = 1.26$). To compensate for this amplification factor by a correction factor, this factor must be $c = \frac{1}{m} = 0.79$.

The offset of the regression is $b = 3.20^\circ$. Please note, that we mentioned in Section 2.2 that the offset for the correction is assumed to be zero when using real measured data ¹⁵²since the calibration of the wind vane should account for such an offset. For the data from the simulation, there is neither ¹⁵³an actual wind vane nor a calibration, and therefore we also get the offset in the numerical simulation results.

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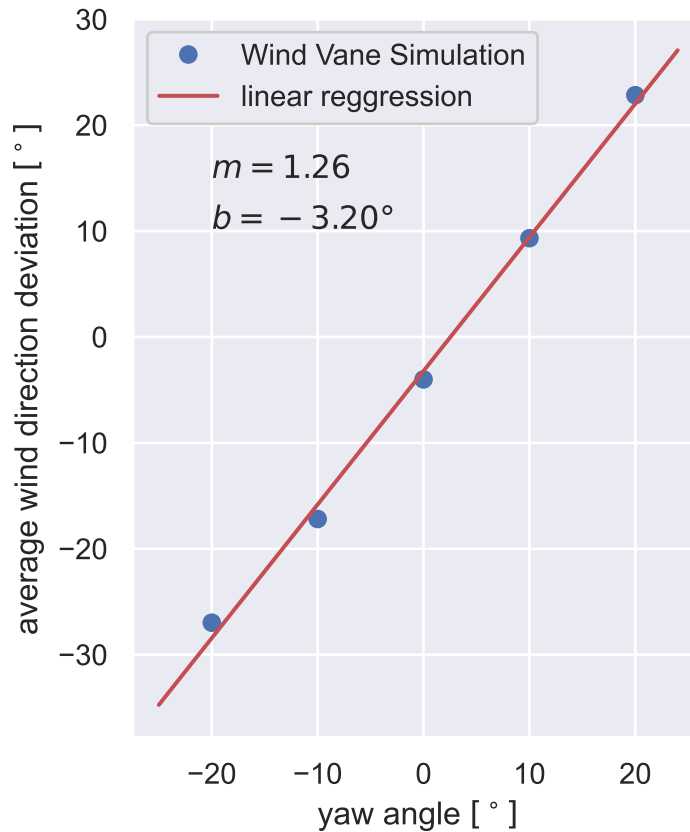


Figure 6. The blue dots represent the average deflection above the nacelle for the five different misalignments from the perspective of the nacelle longitudinal axis, that were simulated in the CFD. In red, a linear regression line was fitted through the points. The slope m and the axis intercept point b are stated.

3.2 Free Field Measurements

295 3.2.1 Comparison Wind Vane versus Met Mast

[..¹⁵⁴] To evaluate the measurements in the free field, we compare the wind direction measurements on the wind turbine nacelle with the measurements at the met mast, as described in Section 2.4.1. First, we present the results from the test field Rysumer Nacken introduced in Section 2.4.1. Here, we only considered wind directions between 180° and 360° , as the met mast is in free flow in this wind direction range and is not disturbed by wakes. The measurements were recorded in the period
 300 from 12.09.2020 to 08.10.2020. After filtering, the amount of viable 60-s measurements is $n = 8013$. [..¹⁵⁵] Figure 7 shows

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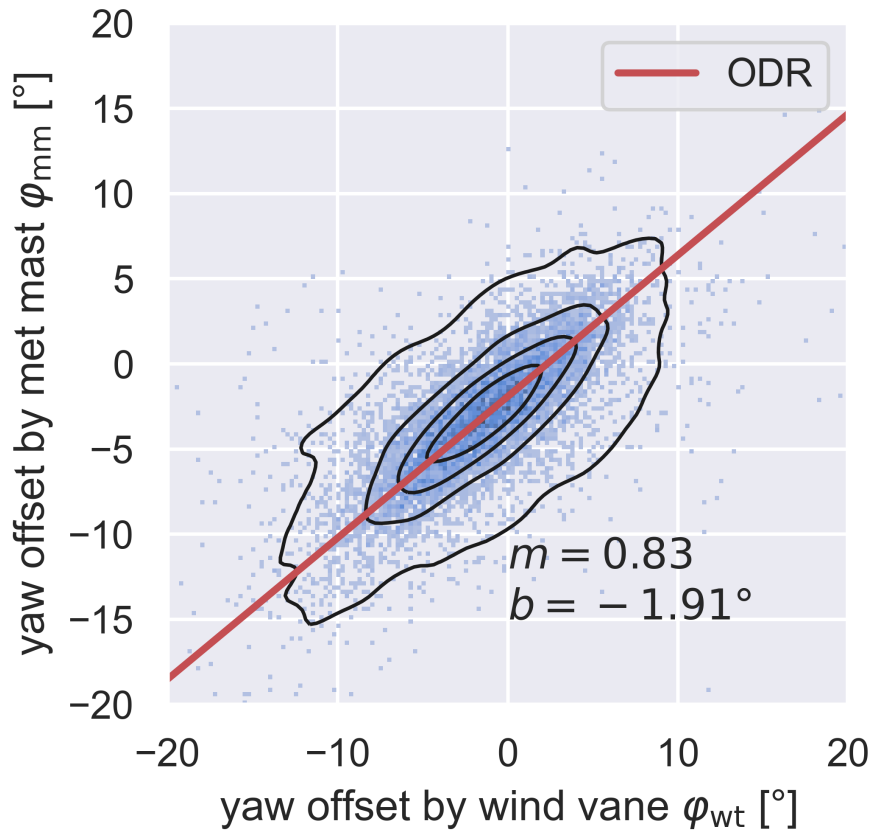


Figure 7. 2D histogram of the yaw misalignment determined by the wind vane at the BARD5.0 φ_{wt} (x -axis) and the met mast φ_{mm} (y -axis) at the Rysumer Nacken. Kernel density estimation is shown in black contour lines and the ODR is displayed in red.

a 2D histogram of the yaw offset measured by the wind vane at the BARD5.0 wind turbine compared to the yaw offset ^[..¹⁵⁶]calculated by met mast.

The black contour lines show a kernel density estimate, and the red line ^[..¹⁵⁷]results from the ODR (see Section 2.4.1). The slope of the regression^[..¹⁵⁸], which also provides the correction factor, is $c := m = 0.83$. The offset is $b = -1.91^\circ$ ^[..¹⁵⁹].

305]. This offset can result from slightly different northings between the wind turbine and the met mast.

Next, we show the results from the test field Kirch Mulsow. In this small wind farm, we only considered wind directions from 215° to 300° , ^[..¹⁶⁰]the sector with free inflow for the met mast, for this evaluation. The measurements were recorded

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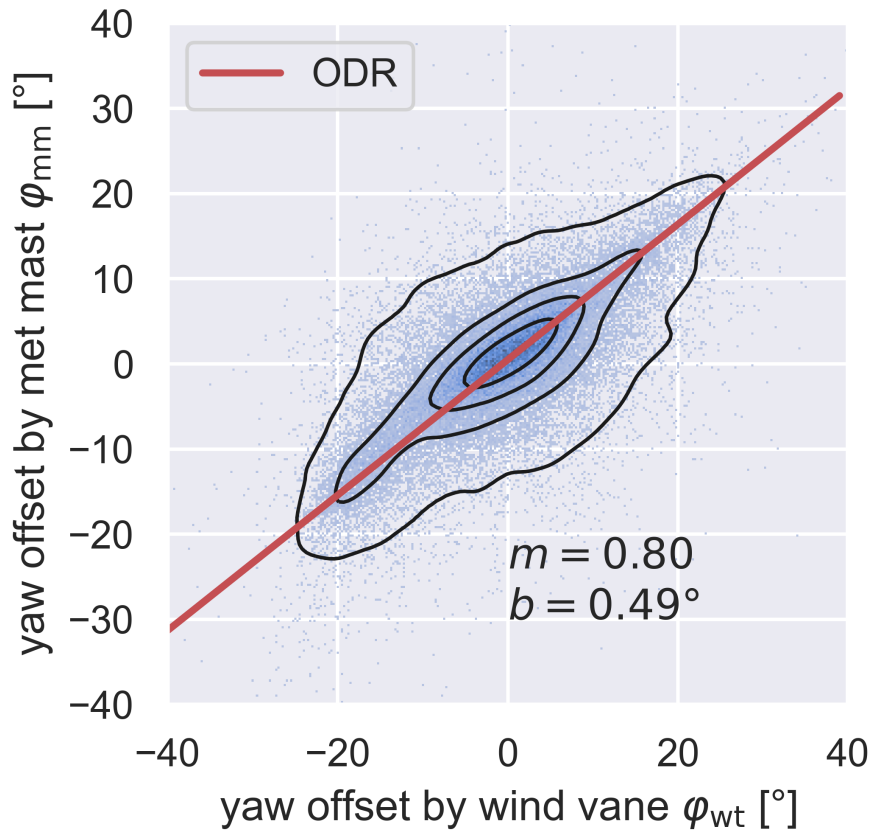


Figure 8. 2D histogram of the yaw misalignment determined by the wind vane at the eno114 φ_{wt} (x -axis) and the met mast φ_{mm} (y -axis) at the Kirch Mulsow site. Kernel density estimation is shown in black contour lines, and the ODR is displayed in red.

in the [\[.161\]](#) period from 19.01.2021 to 03.07.2021. The amount of viable measurements after filtering is $n = 45042$. Figure 8 [\[.162\]](#) compares the measured values of the met mast at the Kirch Mulsow site with those of the wind vane at the eno114 wind turbine in the same way as before. The ODR, in this case, results in a correction factor of $c := m = 0.80$ and the offset is $b = 0.49$.

Since intentional misalignments of the rotor of up to 20° in both directions were also tested at this test site, our range of values here extends much further than in the first case.

3.2.2 Wind Vane Measurements before and after Yaw Actuation

In this section, an analysis of wind direction measurements by the wind vane at the BARD5.0 wind turbine in the Rysumer Nacken test area during the period from 01.08.2020 to 08.10.2020 is performed using the methods presented in section 2.4.2.

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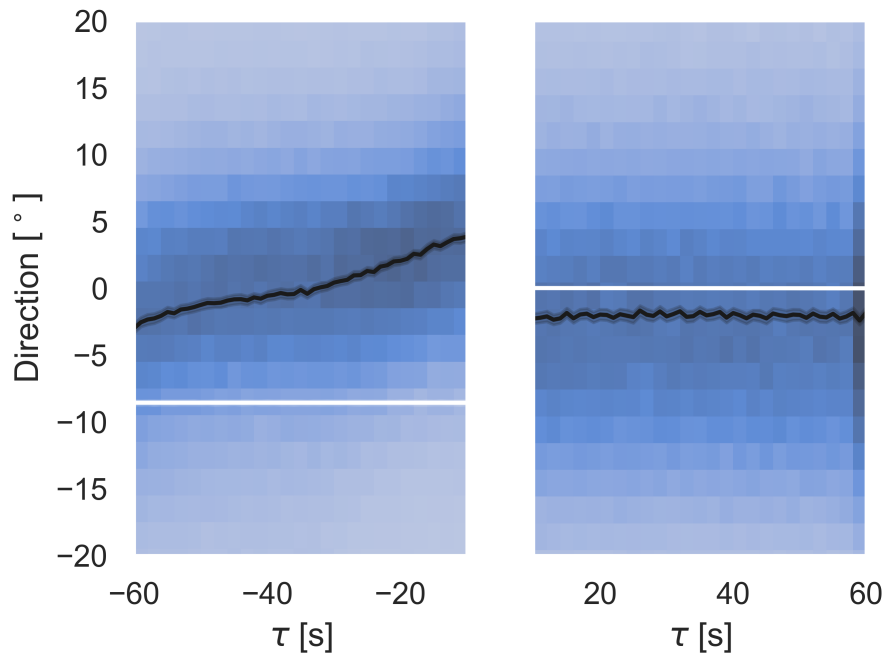


Figure 9. 2D histogram of wind direction measured by the wind vane of the BARD5.0 wind turbine before (left) and after (right) a cw-yaw manoeuvre [..¹⁶³]centred around the yaw angle at the end of the yaw manoeuvre. The blue colouring indicates the number of occurrences, with a darker blue indicating a higher count. The average wind direction $\bar{\omega}$ (see Eq. 4) is shown in black with a 99% confidence interval displayed by the thin grey band around the black line. The yaw angle $\bar{\gamma}$ (see Eq. 5), also with [..¹⁶⁴]its 99% confidence interval, is displayed in white.

g averages. Since the yaw manoeuvre can have an influence on the measured values due to the time averaging, we excluded the measurements 10 seconds before and after the yaw manoeuvre from the analysis. Figure 9 shows the 2D histograms of the wind vane measurements before and after the yaw manoeuvre in cw direction, and Figure 10 the same for yaw manoeuvres in
 320 ccw direction.

In this time period, we recorded 4234 yaw manoeuvres in cw direction and 4082 in ccw direction.

The time series of each yaw [..¹⁶⁷]manoeuvre were averaged as described in Eq. 4 and plotted as a black graph, with [..¹⁶⁸]a 99% confidence interval. The individual measurements of the wind direction are displayed in a 2D histogram, which reflects the frequency by a blue [..¹⁶⁹]colouration. The yaw angle was averaged according to Eq. 5 of the wind turbine is displayed
 325 in white. Since the yaw manoeuvres can be of different magnitudes, we have also given a confidence range for the yaw angle before the manoeuvre, although the confidence range is very small, so [..¹⁷⁰]it is hardly visible.

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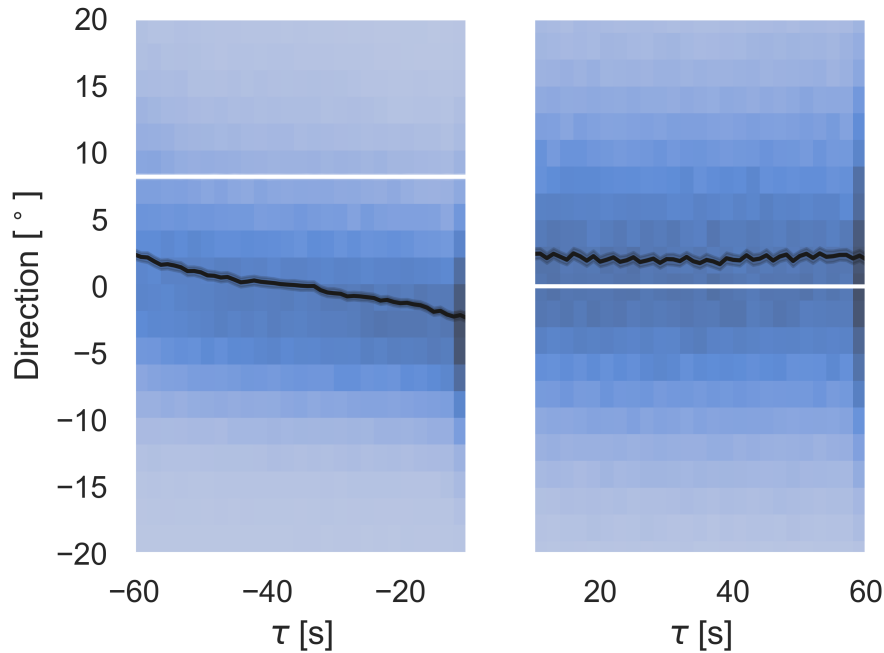


Figure 10. 2D histogram of wind direction measured by the wind vane of the BARD5.0 wind turbine before (left) and after (right) a ccw-yaw manoeuvre [..¹⁶⁵]centred around the yaw angle at the end of the yaw manoeuvre. The blue colouring indicates the number of occurrences, with a darker blue indicating a higher count. The average wind direction $\bar{\omega}$ (see Eq. 4) is shown in black with a 99% confidence interval displayed by the thin grey band around the black line. The yaw angle $\bar{\gamma}$ (see 5), also with [..¹⁶⁶]its 99% confidence interval, is displayed in white.

In both figures, an increase in the wind direction deviation can be seen before the yaw manoeuvre. This increase [..¹⁷¹]in the deviation can be explained by the fact that a yaw manoeuvre is triggered by the yaw controller when a moving average value of the wind direction deviation exceeds a certain threshold value. Since we filtered for exactly these situations, we can see an
 330 increase [..¹⁷²]in the moving average up to the point where a yaw manoeuvre is triggered.

Both figures further reveal that after a yaw manoeuvre, on average, the measured wind direction does not match the orientation of the wind turbine, but that the wind turbine has overshoot the target by 2° to 3° for both cw and ccw yaw directions.

For the calculation of the correction factors, we are using Eq. 11. From the measurements of the cw-yaw manoeuvres we retrieve an average wind direction deviation before the yaw manoeuvre of $\bar{\varphi}_{cw}^{\tau \leq 0} \approx 9.23^\circ$ and after the yaw manoeuvre
 335 of $\bar{\varphi}_{cw}^{\tau > 0} \approx -2.11^\circ$, which results in a correction factor of $c_{cw} = \frac{\bar{\varphi}_{cw}^{\tau \leq 0}}{\bar{\varphi}_{cw}^{\tau \leq 0} - \bar{\varphi}_{cw}^{\tau > 0}} \approx \frac{9.23^\circ}{11.34^\circ} \approx 0.81$. [..¹⁷³]Analogously, the wind

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direction deviation before the yaw manoeuvre for the ccw-cases is $\overline{\varphi}_{ccw}^{\tau \leq 0} \approx -8.28^\circ$ and after the yaw manoeuvre $\overline{\varphi}_{ccw}^{\tau > 0} \approx 2.11$. The correction factor for the ccw-yaw manoeuvres therefor is $c_{ccw} = \frac{\overline{\varphi}_{ccw}^{\tau \leq 0}}{\overline{\varphi}_{ccw}^{\tau \leq 0} - \overline{\varphi}_{ccw}^{\tau > 0}} \approx \frac{-8.28^\circ}{-10.39} \approx 0.80$.

3.3 Free Field Experiment of Wind Vane Correction

We conducted experiments in the test field Rysumer Nacken on the BARD5.0 wind turbine to investigate the effects of wind
 340 vane correction on the operation of a commercial wind turbine. Since the operation of a wind turbine [..¹⁷⁴] depends on uncontrollable and random conditions, we wanted to build a database that would give us a good comparison between normal operation and operation with wind vane correction. For this reason, we ran the wind turbine alternately for one hour in normal operation and one hour with wind vane correction enabled [..¹⁷⁵]. We repeated this procedure until we collected [..¹⁷⁶] sufficient data. We refer to this procedure as a "toggle test".

345 We performed the first toggle test in the period from 06.07.2021 to 26.08.2021. In this experiment, we use the correction model described above (see Eq. (11)) with a correction factor of $c = 0.8$ for the wind vane [..¹⁷⁷] since our investigations at the time of the experiments resulted in this correction factor. This means that the wind vane signal $\varphi_{wt}(t)$, [..¹⁷⁸] used in the yaw controller to steer the wind turbine orientation, was multiplied by this factor directly at the input, i.e. for the yaw trigger [..¹⁷⁹] and the yaw target. We denote the corrected wind vane measurement as $\varphi_{corr}(t) = c \cdot \varphi_{wt}(t)$.

350 In the second toggle test, [..¹⁸⁰] performed from 01.09.2021 to 19.11.2021, we also use a correction factor of $c = 0.8$. [..¹⁸¹] However, this test uses the unmodified wind vane measurement $\varphi_{wt}(t)$ [..¹⁸²] for the yaw trigger. [..¹⁸³] The corrected value $\varphi_{corr}(t)$ is applied only for the yaw target.

For the evaluations of the tests, we consider the number of yaw manoeuvres in the respective test period [..¹⁸⁴]. We analyze the "step response" of the measured wind direction deviation during a yaw manoeuvre according to section 2.4.2, and we
 355 investigate the influence on the power output by calculating the changes in the power curve.

3.3.1 Evaluation of the Toggle Tests

This section summarizes the results of the two toggle tests that were conducted in the Rysumer Nacken test site at the BARD5.0 wind turbine. Table 1 lists the most important statistics of the toggle tests, which are referred to in the following when describing the individual test results.

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Table 1. Results of the toggle tests at Rysumer Nacken.

| statistic | Toggle Test 1 | | Toggle Test 2 | |
|---------------------------|------------------|------------------|------------------|------------------|
| | Normal operation | WVane Correction | Normal operation | WVane Correction |
| total duration | 21.07 days | 20.95 days | 31.93 days | 32.52 days |
| number of yaw actuations | 4960 | 2869 | 5373 | 5047 |
| yaw actuations per 10 min | 1.64 | 0.95 | 1.17 | 1.08 |
| total yaw distance | 41450° | 21954° | 42625° | 31522° |
| yaw distance per 10 min | 13.66° | 7.28° | 9.27° | 6.73° |

360 Even though the period of normal operation was the same as the period with activated wind vane correction when performing a toggle test, the "total durations" differ [..¹⁸⁵]here. The data filtering can explain this, since only periods were considered in which the wind turbine was active, operated in the partial load range, and no active power curtailment was applied.

In the first toggle test, the turbine performed on average 0.95 yaw actuations in 10 minutes with the activated wind vane correction, turning 7.28° per 10 min on average. Compared to 1.64 yaw manoeuvres and a yaw distance of 13.66° per 10 min
365 on average in normal operation, this is a reduction of 41.8% in actuations and 46.7% in total yaw distance.

The yaw distance, in general, is closely related to the actuator duty cycle (ADC) of the yaw controller. The BARD5.0 turbine yaws at an average rotational speed of about 0.75° per second, so the average yaw distance per 10 min of 13.66° for normal operation takes approximately 18.21 seconds, which means that the yaw motor is active about 3% of the time. Due to the inertia of the nacelle, the rotation speed is not entirely constant; therefore, this conversion is only an
370 approximation. Thus, in this evaluation, we only use the average yaw distance per 10 min and not additionally the ADC.

In the second toggle test, the yawing activity was reduced from 1.169 yaw actuations per 10 minutes in regular operation to 1.078 per 10 minutes with wind vane correction activated, i.e. by 7.8%. [..¹⁸⁶]Due to the wind vane correction, the yaw distance experienced a reduction of 27.4% from 9.27° to 6.73° per 10 minutes[..¹⁸⁷].

For both test cases, we aggregated the wind direction deviation before and after the yaw processes as described in Section
375 2.4.2 and present them in Figures 11 and 12 for the first test case analog to Figure 9 and Figure 10 . The measurements shown are the 10-s moving average wind direction measurements. It can be seen that the alignment of the wind turbine and the measured wind direction at the wind vane match better after the yaw process compared to Figure 9 and Figure 10. The result for the second test case looks very similar to this figure, so we do not show it here.

To identify the influence of the wind vane correction on the performance, we determined the power curves binned in 1 m/s
380 steps from the 10-min data for both conditions in both test cases[..¹⁹²]. We calculated the absolute power difference for each

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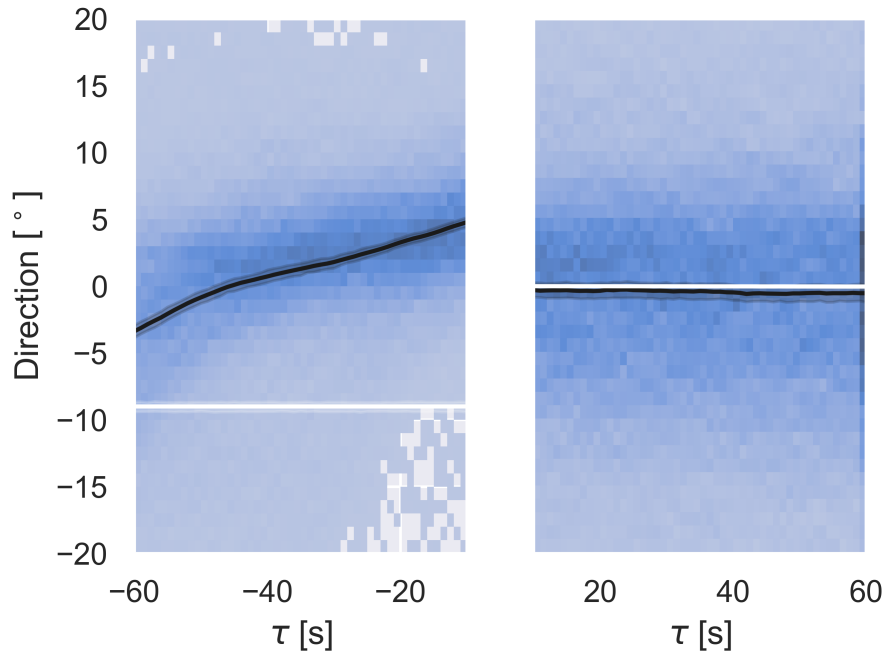


Figure 11. 2D Histogram of wind direction measured by the wind vane of the BARD5.0 before (left) and after (right) a cw-yaw manoeuvre [..¹⁸⁸]centred around the yaw angle at the end of the yaw manoeuvre for activated wind vane correction in the first toggle test. The average wind direction and its 99 % confidence interval [..¹⁸⁹]are shown in black. The yaw angle is displayed in white.

bin $P_{\text{diff}} = P_{\text{corr}} - P_{\text{standard}}$, where P_{corr} is the power output with the wind vane correction active and P_{standard} is the power output during standard yaw control, with no wind vane correction.

Figure 13 shows the difference of these power curves for the first toggle test and Figure 14 for the second toggle test. In addition to the difference, we have added error bars. These were calculated from the square root of the sum of the squared standard errors of the mean of both power curves multiplied by 2.576 to give an estimate of the 99 % confidence interval.

Figure 13 shows that in the first toggle test [..¹⁹⁹], on average, less power is produced by the wind vane correction for most wind speeds [..²⁰⁰]. Still, the fluctuations in the measurements are so large that this is not statistically significant. In Figure 14, more power is produced during the second toggle test for most wind speeds by the wind vane correction, but the difference is not statistically significant here either.

For both test cases, we calculated the influence on the Annual Energy Production (AEP) using the average values and an assumed Weibull wind distribution (Weibull scale parameter $A = 11.33$ m/s and Weibull shape parameter $k = 2.29$). For toggle test 1, this results in a loss of power generation of -0.43% and for toggle test 2 an increase in power of 0.06% .

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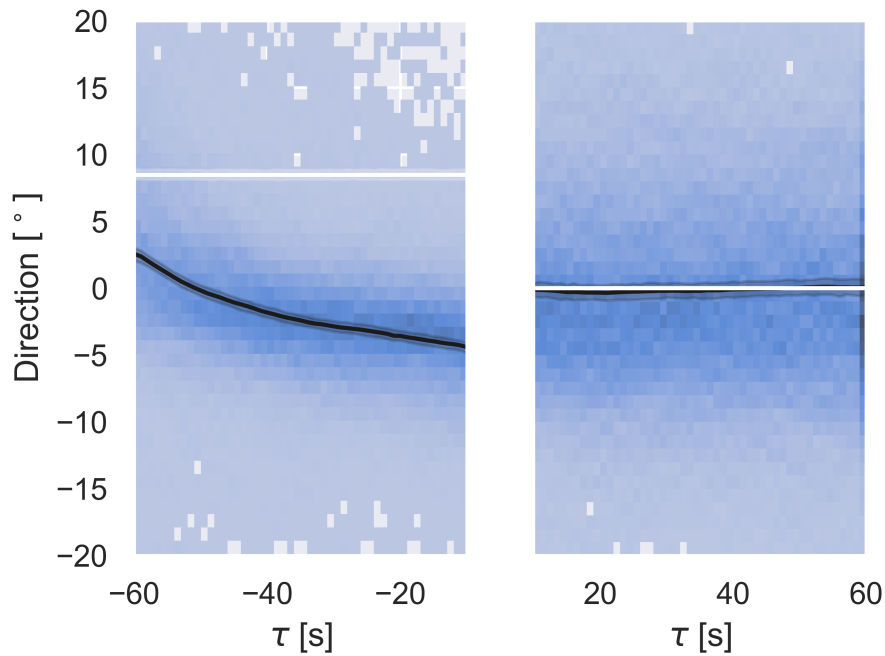


Figure 12. 2D Histogram of wind direction measured by the wind vane of the BARD5.0 before (left) and after (right) a ccw-yaw manoeuvre [\[..¹⁹⁰\]](#) centred around the yaw angle at the end of the yaw manoeuvre for activated wind vane correction in the first toggle test. The average yaw misalignment and the 99 % confidence interval [\[..¹⁹¹\]](#) are shown in black. The yaw angle is displayed in white.

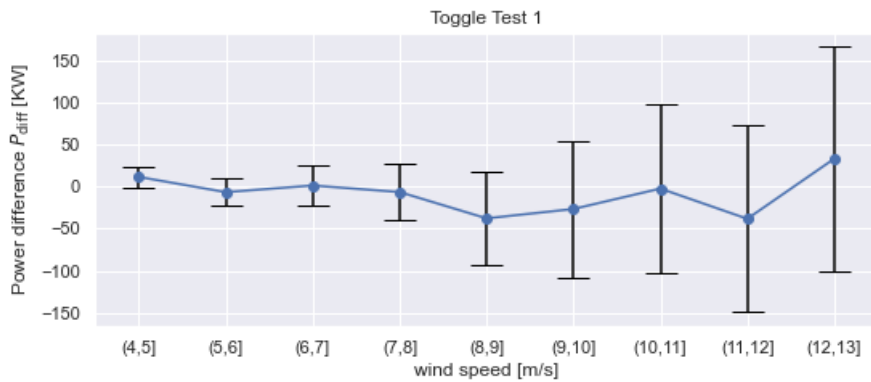


Figure 13. [\[..¹⁹³\]](#) The power difference between activated wind vane correction and regular operation over wind speed [\[..¹⁹⁴\]](#) binned in 1 m/s, with error bar, which [\[..¹⁹⁵\]](#) represents the 99% confidence interval for Toggle Test 1.

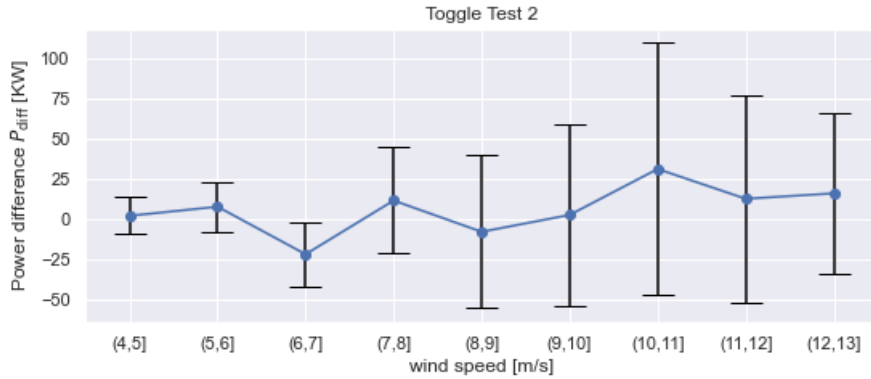


Figure 14. [..¹⁹⁶]The power difference between activated wind vane correction and regular operation over wind speed [..¹⁹⁷]binned in 1 m/s, with error bar, which [..¹⁹⁸]represents the 99% confidence interval for Toggle Test 2.

4 Discussion

The CFD simulations we performed (see Section 3.1) to better understand the mean wind direction immediately behind the rotor plane, support our hypothesis that the rotor’s thrust deflects the flow at the wind vane location during yaw misalignment. This effect affects the far wake, as previously shown in other studies (Jiménez et al., 2010; Bastankhah and Porté-Agel, 2016), and has an impact on the flow directly above the nacelle, and thus, on conventional wind vanes of a wind turbine. However, our simulations only serve as a proof-of-concept, as we used ideal conditions (uniform flow) for the calculations in order to be able to represent the effect of the rotor on the flow in isolation. We focused our investigation on an average error in the wind vane signal. For modelling the wind vane error, we have assumed a simple affine linear function, which seems to be confirmed in the CFD simulations in the range of -20° to 20° . In general, we believe that this model is appropriate for small misalignments but no longer applies for larger misalignments, as the turbine’s thrust decreases in these situations and the overestimation of the wind vane should not increase linearly anymore. The influence of more complex conditions, such as increased turbulence intensity as in an unstable atmospheric condition or stronger veer and shear in the flow as in a stable atmospheric condition, was not investigated in the simulation here. For further future investigations, spatially high-resolution LES simulations could be used to study complex conditions and dynamic inflows in combination with different yaw controllers.

The comparisons between met mast measurements and wind turbine measurements (see Section 3.2.1) also show that the wind vane tends to overestimate the deviation of the wind direction, both at the BARD5.0 wind turbine at the Rysumer Nacken test site and [..²⁰¹]the eno114 wind turbine in Kirch Mulsow. A direct comparison is [..²⁰²]difficult because the met mast is located at [..²⁰³]about $2.5 D$ (approx. 300 m) from the wind turbine and [..²⁰⁴], therefore, experiences a slightly different

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flow. In addition, we have filtered out wind direction sectors for each site, as in these, the met mast would be in the wake of the surrounding wind turbines, and the wind direction measurement would thus be affected. Nevertheless, [..²⁰⁵] the measurements on both test sites confirm our hypothesis on average. In the case of the Rysumer Nacken test field, it can be observed that the measurement data is not centred around the 0°. This is due to the fact that the measurement values of the wind vane are
415 the raw values without the offset correction and that the measurement mast has a slightly different northing to the orientation of the wind turbine. [..²⁰⁶] We received the offset-corrected measured values at the Kirch Mulsow test site. However, in the case of Kirch Mulsow it is also interesting that we have considered larger misalignments here, and it shows in Figure 8 that the values in the centre [..²⁰⁷] receive a proportionally larger deflection than the outer values, which speaks against a linear model for the overestimation. One can assume that the thrust decreases with greater misalignment, and therefore, the deflection also
420 becomes weaker and no longer increases linearly. For the misalignments analysed here, however, the linear model seems to be sufficiently accurate. In a similar study, Simley et al. (2021) compared wind vane measurements with those obtained from a nacelle-based lidar, in contrast to our [..²⁰⁸] research, which utilised measurements from the met mast. Although Simley conducted his study on a different turbine and employed [..²⁰⁹] other reference measurements, his findings are consistent with [..²¹⁰] ours, demonstrating that the wind vane overestimates wind direction deviation.

425 Simley also encountered a problem with regression dilution when performing linear regression. To mitigate this issue, he attempted to reduce the uncertainties in the prediction variable through binning. In contrast, we opted to use orthogonal distance regression (ODR), which accounts for the uncertainties in both variables, to circumvent the problem with linear regression.

The comparison of the wind vane measurements before and after the yaw manoeuvre (Section 3.2.2) has the advantage that it can be performed without external measurements [..²¹¹]. Thus, a correction factor for the wind vane can be determined for
430 a wind turbine [..²¹²] with no reference measurement, such as a measuring mast or a lidar. Similar to the analysis of a step response for linear time-invariant control systems, this method opens up the possibility of analysing the effect of misalignment on a variety of variables, such as power, wind speed measurements or load measurements, if available, albeit in an empirical rather than deterministic manner.

The evaluations of both toggle tests (Section 3.3.1) have shown that the correction factor improved the wind turbine's
435 alignment in the wind direction on average after the yawing process. In the first toggle test, the correction factor was applied to the wind vane measurement for the yaw trigger [..²¹³] and the yaw target. The results show [..²¹⁴] that the yawing activity was significantly reduced (by 41.82%), but this was at the expense of performance, as the corrected wind vane resulted in larger misalignments than in normal operation. However, the reduction of yawing activity is not only related to allowing larger

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misalignments, as shown by the evaluation of the second toggle test, where the yaw trigger is the same as in normal operation, only the yaw target is affected by the corrected wind vane signal. Nevertheless, the yawing activity was still reduced by 7.78 % compared to normal operation due to a better alignment of the wind turbine after the yaw manoeuvre. Especially the number of alternating yaw manoeuvres could be significantly reduced. And since the yaw distance of each yaw manoeuvre was by definition 20 % shorter in the second toggle test, the total yaw distance was reduced by more than 27 %. The performance could ^[..²¹⁵]increase slightly in the second toggle test due to the better alignment. ^[..²¹⁶]However, the decrease in power on the first toggle test and the increase in power on the second ^[..²¹⁷]test are not statistically significant. ^[..²¹⁸]More extended test periods are needed to determine this. ^[..²¹⁹]Mainly because the measurements are not independent samples but correlated time series. The effective sample size is ^[..²²⁰], therefore, even smaller.

Table 1 reveals that the time period of the tests has a large influence on the number of yaw manoeuvres. We suspect that during the second toggle test, which took place in autumn, stable atmospheric stratification was more frequent compared to the first toggle test in summer, and that this had an influence on the number of yaw manoeuvres in general. Therefore, it made sense to conduct toggle tests, as the conditions for both test states were largely identical. Nevertheless, a longer test period and a differentiation of the effects of the wind vane correction for different atmospheric stabilities could provide further useful insights.

In our study, each of our investigations (CFD simulation in Section 3.1, wind vane to met mast comparison in Section 3.2.1, yaw manoeuvre analysis (Section 3.2.2)) resulted in a correction factor of approximately 0.8 and the toggle tests in Section 3.3.1 shows great improvement for applying a wind vane correction with this factor. This indicates the general order of magnitude for the correction factor, but this value will depend on the shape of the turbine nacelle and the placement of the wind vane on the nacelle. Therefore, the authors advise that the methods presented here be carefully repeated using data from a wind turbine before applying a correction factor to other wind turbine types.

Overall, our results are consistent with those of (Simley et al., 2021) and (Kragh and Fleming, 2012), indicating that for yaw control based on wind vanes, transfer functions ^[..²²¹]and offset calibration are required to correct wind vane overshoot. In our investigation, we presented methods for ^[..²²²]parameterising the transfer function for the specific wind turbine. Atmospheric stability can have a significant impact on these parameters. Therefore, in future studies, we will conduct analyses to determine atmospheric stability and filter data accordingly.

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Our study on wind vane measurements during yaw misalignment on two commercial wind turbines and a CFD simulation revealed that wind direction deviation was overestimated by about 20 % to 30 %, with the CFD simulation supporting the hypothesis that this is an inherent characteristic due to the rotor's thrust. To mitigate this problem, we developed a linear correction function and two data-driven methods to parameterize it. These methods involved using measurements from a meteorological mast or ²²³analysing wind direction measurements before and after yaw manoeuvres. We tested the correction function on one wind turbine in two scenarios. In the first scenario, the correction was applied to both the yaw trigger ²²⁴and the yaw target, reducing yaw activity by more than 40 %. In the second scenario, the correction was only applied to the yaw target, resulting in a reduction of yaw activity by approximately 8 %. Results also indicated an improvement in alignment with the flow in both scenarios, while power production of the wind turbine was not significantly influenced. Our ²²⁵findings suggest that a corrected wind vane signal is crucial for improved wind turbine control strategies, particularly for wake deflection²²⁶. Future studies could explore ²²⁷using our correction function and data-driven methods on other wind turbine types and in different environmental conditions.

Author contributions. AR coordinated the research, developed the methods, and performed the analysis of the measured data. LH performed the CFD simulation, provided the simulation data, and assisted with valuable discussions. PH aided with helpful discussions and supported in gaining access to and an overview of the data from the Kirch Mulsow site. LJL provided important advice regarding the simulations and provided thorough internal review and valuable feedback. CM was responsible for providing data from the Rysumer Nacken test site, integrating the correction factor into the BARD5.0 turbine controller, performing the toggle tests, and providing important advice on methods and data analysis. MK provided thorough internal review and valuable feedback and had a supervisory function.

Competing interests. The authors declare that they have no conflict of interest.

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References

- 490 Bastankhah, M. and Porté-Agel, F.: Experimental and theoretical study of wind turbine wakes in yawed conditions, *Journal of Fluid Mechanics*, 806, 506–541, <https://doi.org/10.1017/jfm.2016.595>, 2016.
- Bertelè, M., Bottasso, C. L., Cacciola, S., Adegas, F. D., and Delpont, S.: Wind inflow observation from load harmonics, *Wind Energy Science*, 2, 615–640, <https://doi.org/10.5194/wes-2-615-2017>, 2017.
- Boggs, P. T. and Rogers, J. E.: Orthogonal distance regression, *Contemporary Mathematics*, 112, 183–194, 1990.
- 495 Bromm, M., Rott, A., Beck, H., Vollmer, L., Steinfeld, G., and Kühn, M.: Field investigation on the influence of yaw misalignment on the propagation of wind turbine wakes, *Wind Energy*, pp. 1–18, <https://doi.org/10.1002/we.2210>, 2018.
- Burton, T., Jenkins, N., Sharpe, D., and Bossanyi, E.: *Wind Energy Handbook Second Edition*, John Wiley & Sons, Ltd, <https://doi.org/10.1002/9781119992714>, 2011.
- Frost, C. and Thompson, S. G.: Correcting for regression dilution bias: comparison of methods for a single predictor variable, *Journal of the Royal Statistical Society: Series A (Statistics in Society)*, 163, 173–189, <https://doi.org/10.1111/1467-985x.00164>, 2000.
- 500 Gebraad, P. M. O., Teeuwisse, F. W., van Wingerden, J. W., Fleming, P. A., Ruben, S. D., Marden, J. R., and Pao, L. Y.: Wind plant power optimization through yaw control using a parametric model for wake effects-a CFD simulation study, *Wind Energy*, 19, 95–114, <https://doi.org/10.1002/we.1822>, 2016.
- Held, D. P. and Mann, J.: Detection of wakes in the inflow of turbines using nacelle lidars, *Wind Energy Science*, 4, 407–420, <https://doi.org/10.5194/wes-4-407-2019>, 2019.
- 505 Hulsman, P., Sucameli, C., Petrović, V., Rott, A., Gerds, A., and Kühn, M.: Turbine power loss during yaw-misaligned freefield tests at different atmospheric conditions, *Journal of Physics: Conference Series*, 2265, 032 074, <https://doi.org/10.1088/1742-6596/2265/3/032074>, 2022.
- International Electrotechnical Commission: *Wind energy generation systems - Part 12-3: Power performance - Measurement based site calibration*, IEC Standard, iEC 61400-12-3, 2022.
- 510 Jeong, J. and Hussain, F.: On the identification of a vortex, *Journal of Fluid Mechanics*, 285, 69–94, 1995.
- Jiménez, Á., Crespo, A., and Migoya, E.: Application of a LES technique to characterize the wake deflection of a wind turbine in yaw, *Wind Energy*, 13, 559–572, <https://doi.org/10.1002/we.380>, 2010.
- Jonkman, J., Butterfield, S., Musial, W., and Scott, G.: *Definition of a 5-MW Reference Wind Turbine for Offshore System Development*, Tech. rep., NREL, <https://doi.org/10.2172/947422>, 2009.
- 515 Kragh, K. and Fleming, P.: Rotor Speed Dependent Yaw Control of Wind Turbines Based on Empirical Data, in: *50th AIAA Aerospace Sciences Meeting including the New Horizons Forum and Aerospace Exposition*, p. 1018, <https://doi.org/10.2514/6.2012-1018>, 2012.
- Mittelmeier, N. and Kühn, M.: Determination of optimal wind turbine alignment into the wind and detection of alignment changes with SCADA data, *Wind Energy Science*, 3, 395–408, <https://doi.org/10.5194/wes-3-395-2018>, 2018.
- 520 OpenFOAM: *Open Source Field Operation and Manipulation*, <https://openfoam.org/>, last accessed 21 October 2021, 2021.
- Pedersen, T., Demurtas, G., and Zahle, F.: Calibration of a spinner anemometer for yaw misalignment measurements, *Wind Energy*, 18, 1933–1952, <https://doi.org/10.1002/we.1798>, 2014.
- Rahimi, H., Daniele, E., Stoevesandt, B., and Peinke, J.: Development and application of a grid generation tool for aerodynamic simulations of wind turbines, *Wind Engineering*, 40, 148–172, <https://doi.org/10.1177/0309524x16636318>, 2016.

- 525 Rott, A., Doekemeijer, B., Seifert, J. K., van Wingerden, J.-W., and Kühn, M.: Robust active wake control in consideration of wind direction variability and uncertainty, *Wind Energy Science*, 3, 869–882, <https://doi.org/10.5194/wes-3-869-2018>, 2018.
- Schreiber, J., Bottasso, C. L., and Bertelè, M.: Field testing of a local wind inflow estimator and wake detector, *Wind Energy Science*, 5, 867–884, <https://doi.org/10.5194/wes-5-867-2020>, 2020.
- Simley, E., Fleming, P., Girard, N., Alloin, L., Godefroy, E., and Duc, T.: Results from a wake-steering experiment at a commercial wind plant: 530 investigating the wind speed dependence of wake-steering performance, *Wind Energy Science*, 6, 1427–1453, <https://doi.org/10.5194/wes-6-1427-2021>, 2021.
- Spalart, P. R., Deck, S., Shur, M. L., Squires, K. D., Strelets, M. K., and Travin, A.: A New Version of Detached-eddy Simulation, Resistant to Ambiguous Grid Densities, *Theoretical and Computational Fluid Dynamics*, 20, 181–195, <https://doi.org/10.1007/s00162-006-0015-0>, 2006.
- 535 Teubler, J.: Ökologische Betrachtung eines Offshore-Windparks unter Berücksichtigung der Netzintegration mit einer Hochspannungsgleichstromübertragung, Ph. D. Thesis, 2011.