



Turbulence Distortion Matters in Predicting Inflow Turbulence Noise of Future Wind Turbines

Özgür Yalçın¹, Andrea Piccolo¹, Riccardo Zamponi^{1,2}, Daniele Ragni¹, and Roberto Merino-Martinez³

Abstract. This study examines the role of turbulence distortion in predicting inflow turbulence (IT) noise generation from large wind turbines via Amiet's theory. Two subsequent distortion mechanisms are investigated: (i) the streamtube expansion in the rotor induction zone and (ii) the interaction with the surface of thick blade profiles. Large-eddy simulations reveal that the turbulence spectra, which reflect distortion effects, remain largely unaffected by rotor induction within the frequency range relevant for noise generation. As for the other mechanism, the distortion of the turbulence approaching a blade leading edge is modeled with a simplified closed-form solution of Goldstein's Rapid Distortion Theory. This vorticity-deflection-based model is extended here beyond the high-frequency approximation and integrated into an analytical Amiet-based IT noise tool. Applications to representative test cases show that while distortion effects are minimal for current turbine sizes, they become relevant for future configurations featuring larger rotor sizes and thicker airfoils. The developed model reveals that IT noise levels do not necessarily scale with rotor size, but are shaped by spectral changes induced by the blade geometry, operational parameters, and inflow conditions. This model offers a physically consistent, computationally efficient framework for aeroacoustic assessment of next-generation wind turbine design.

Nomenclature

 (∞) denotes undistorted variable

 (X_1, X_2, X_3) drift coordinates

 (x_1, y_1, z_1) translated reference frame

 (x_2, y_2, z_2) rotated reference frame

 (x_a, y_a, z_a) Amiet-oriented reference frame

 (x_h, y_h, z_h) hub-based reference frame

20 α local angle of attack

¹Department of Flow Physics and Technology, Delft University of Technology, Kluyverweg 1, 2629HS Delft, the Netherlands ²Environmental and Applied Fluid Dynamics Department, von Karman Institute for Fluid Dynamics, Waterloosesteenweg 72, B-1640 Sint-Genesius-Rode, Belgium

³Department of Control and Operations, Delft University of Technology, Kluyverweg 1, 2629HS Delft, the Netherlands **Correspondence:** Özgür Yalçın (o.e.yalcin@tudelft.nl)







30

35

40

 θ

angle of r_o relative to z_h -axis

 β local pitch angle β_0 blade pitch angle κ distorted wavenumber domain $u^{(g)}$ gust velocity Umean velocity unsteady velocity vlocal azimuth angle γ denotes complex amplitude \hat{s} unit direction vector from the retarded source to the observer \mathcal{L} aeroacoustic transfer function Ω blade angular velocity Doppler-shifted frequency ω emission frequency ω_e ϕ_{11} streamwise energy spectrum upwash energy spectrum ϕ_{22} angle of M_t relative to y_h -axis ψ freestream density ρ_{∞} amplified distance σ σ_s distance from the upstream along the streamline AR area ratio SPL sound pressure level ΤI turbulence intensity Θ angle between M_c and \boldsymbol{x}_o





velocity potential term 45 \boldsymbol{x} observer position relative to local Amiet frame observer position relative to hub frame \boldsymbol{x}_o \boldsymbol{x}_s present source position relative to hub frame dummy spatial coordinate ξ magnitude of the gust velocity 50 aaxial induction factor a_x Bnumber of blades span of a segment chord of a segment c55 c_{∞} speed of sound D/Dt substantial derivative Drotor diameter D_{ij} distortion tensor f frequency 60 kwavenumber Lincoming turbulent length scale distortion length scale $l_{
m dist}$ L_w sound power level l_y correlation length Mincoming Mach number aligned with local Amiet x-axis 65 M_b rotational blade Mach number M_c convection Mach number M_s azimuthal Mach number





	M_t	transversal freestream Mach number
70	M_z	axial freestream Mach number
	M_{rel}	relative Mach number between the incoming flow and the moving segment
	$N_{ m seg}$	number of segments per blade
	Q	magnitude of the velocity potential
	R	cylinder/rotor radius
75	r	radial location of a segment
	r_e	distance between the observer and the retarded source position
	r_o	direct observer distance from the hub
	R_y	rotation matrix about y -axis
	R_z	rotation matrix about z-axis
80	S_{22}	power spectral density of upwash velocity fluctuations
	S_{pp}'	radiated power spectral density at an azimuthal angle
	U	mean flow speed
	u	fluctuating flow speed
	U_{∞}	average wind/freestream speed
85	u_{∞}	fluctuating freestream speed
	U_x	incoming velocity in x-axis

1 Introduction

 u_x

incoming velocity fluctuation in x-axis

The recent increased demand for renewable energy has promoted the deployment of new, larger wind turbines (WTs) (Global Wind Energy Council, April 2024; Wind Europe, February 2025). One of the major concerns for onshore WTs, however, remains their noise emissions (Kirkegaard et al., 2025), which restrict their placement close to residential areas due to national noise regulations (Davy et al., 2018). In modern WTs, mechanical noise has been significantly reduced thanks to advances in gearbox technology, making aerodynamic noise the predominant source (Rogers and Manwell, 2004; Oerlemans et al.,



100

105

110

115

120

125



2007). Aerodynamic noise can be classified mainly into blade self-noise and inflow turbulence (IT) noise. Considering blade optimization and operational flow regimes, trailing edge (TE) noise is a dominant self-noise source, which inherently scatters in the mid-to-high frequency range (Oerlemans et al., 2007). On the other side, IT noise arises from the interaction between the incoming turbulent flow and the blade leading edges (LEs), predominantly emitting low-frequency noise ($f < 200 \, \mathrm{Hz}$, corresponding to acoustic wavelengths larger than blade chords and comparable to large WT rotor scales) (Lowson, 1993; Buck et al., 2016).

Recent developments, such as TE add-ons and active flow control mechanisms, have yielded notable reductions in TE noise (Zhou et al., 2022; Lahoz et al., 2024). However, increasing rotor sizes and hub heights of next-generation turbines shifts the noise spectrum to lower frequencies, thus amplifying the relevance of IT noise (Møller and Pedersen, 2011). This shift is particularly important considering that low-frequency sound waves propagate over longer distances and experience reduced air absorption and transmission losses by building materials. Therefore, they are more perceptible indoors due to room resonances, which increase their negative health impact (Blumendeller et al., 2020). Consequently, accurate prediction of IT noise has become as critical as TE noise for the development of quieter WTs.

Several low-fidelity approaches for predicting IT and TE noise in rotational motion (Moriarty and Migliore, 2003; Sinayoko et al., 2013; Tian et al., 2013; Buck et al., 2016; Botero-Bolívar et al., 2024) have been developed as extensions of Amiet's theory (Amiet, 1975), which was initially proposed for rectilinear motion. These methods represent an efficient approach in terms of cost and accuracy, particularly for IT noise predictions, since the relevant flow region just upstream of the blade LE is minimally affected by the boundary layer development. Indeed, Amiet's model, originally formulated for a flat plate with a relatively large span immersed in a subsonic turbulent flow, relates the far-field acoustic power spectral density to the incoming gust velocity spectrum via a transfer function. This means that IT characteristics are the only required inputs to obtain an estimate of the acoustic performances.

A primary limitation of Amiet's model lies, therefore, in its dependence on the IT spectrum, which is inherently difficult to predict and measure. In practice, the inflow is assumed to be isotropic and characterized by canonical spectra, such as von Kármán or Liepmann models, parameterized by upstream turbulence intensity and integral length scale. However, experimental and numerical/analytical studies have demonstrated that the upstream turbulence distorts as it approaches the LEs of thick airfoil sections (Batchelor and Proudman, 1954; Lighthill, 1956; Hunt, 1973; Paterson and Amiet, 1976), resulting in an alteration of its velocity spectrum. The key mechanism has been identified as the distortion of the vorticity field in the stagnation region, which reduces the upwash velocity in the wavenumber range associated with turbulent scales smaller than the characteristic length (e.g., LE radii). This distortion leads to a marked reduction in the slope of the upwash velocity spectrum at high frequencies, resulting in a corresponding attenuation of noise emissions in the same range (Atassi et al., 1990; Moreau and Roger, 2005; Mish and Devenport, 2006; Christophe, 2011). On the one hand, this highlights the influence of airfoil thickness on sound generation; on the other, it demonstrates that neglecting this effect can lead to discrepancies between predicted and measured IT noise levels (Devenport et al., 2010; Santana et al., 2016; Buck et al., 2018; Zamponi et al., 2020; Piccolo et al., 2023).In particular, it is important to use the velocity spectrum sampled or modeled as close to the LE as possible, as input to Amiet's model.



130

135

140

145

150

155

160



To address this limitation, several enhancements have been proposed to incorporate turbulence distortion and thickness effects (Moreau and Roger, 2005; Santana et al., 2016; dos Santos et al., 2023; Dos Santos et al., 2024), also for WTs, considering their relatively thick blade profiles (Moriarty et al., 2005; Buck et al., 2018; Faria et al., 2020; Botero-Bolívar et al., 2024). These modifications typically adjust empirical parameters within the canonical turbulence spectrum, particularly targeting its inertial subrange to better capture variations in spectral slope and hence improve IT noise predictions. Some studies have also explored these adaptations for rotational motion, such as using one-dimensional turbulence spectra to integrate single-hot-wire measurements (Piccolo et al., 2025).

While these corrected models have shown better agreement with experimental data in wind tunnel flows over moderately thick airfoils, their applicability to full-scale WT operations, where incoming turbulence has integral length scales much larger than the airfoil chord, is questionable (Faria et al., 2020; Botero-Bolívar et al., 2024). These models may underpredict IT noise under atmospheric turbulence conditions, where canonical models sometimes perform better (Botero-Bolívar et al., 2024).

Alternatively, the Rapid Distortion Theory (RDT), that is based on linearized vorticity equations, can represent a valid approach to account for the distortion effects, as it allows for a more generalized and realistic framework. In principle, it models the distortion of weak turbulent perturbations under mean flow convection conditions in cases where the coherence between eddies is unaltered. Originally developed by Batchelor and Proudman (1954) and extended by Hunt (1973), the RDT formulates distortion through a set of coupled partial differential equations to describe the upstream field of a two-dimensional bluff body by wavenumber analysis. It distinguishes two mechanisms according to the relative size of the eddies with respect to a characteristic geometric dimension of the body. The first is the blockage due to the body surface, and the second is the vorticity deflection due to the deformation of the mean flow streamlines.

Although the asymptotic results of the RDT have already been incorporated into Amiet's model as corrections in the mentioned studies, Zamponi et al. (2021) proposed a way to implement the full solution of Hunt's formulation. The resulting estimation of distortion effects could support the development of a tool for assessing the performance of noise-reduction technologies. However, the considerable complexity and high computational cost of the implementation restricted the wavenumber range over which solutions could be obtained, limiting the practicality of this methodology for low-fidelity applications.

An alternative RDT approach, originally proposed by Goldstein (1978), is based on a linearized Euler equation for non-uniform, homentropic mean flows. Solving these equations using a Lagrangian framework in drift coordinates allows for direct tracking of turbulent flow in the distorted field. This method is especially suitable for low-speed flows, such as those typical for the upstream regions of the WT blade profiles. For high-frequency gusts, a simplified version of this model was later derived by Majumdar and Peake (1998) to estimate fan inflow noise.

Turbulence distortion is not only related to the thickness of the aerodynamic object in the convective direction. In wind energy applications, the expansion of the streamtube induced by the WT rotor also plays a role by creating a region of flow deceleration. Unlike contracting flows in wind tunnels or propulsion systems, streamtube expansion is associated with non-uniform and inhomogeneous strain-rate fields in the induction zone, violating key assumptions of considering distortion as rapid (Batchelor and Proudman, 1954). As a result, the distortion of large energetic eddies, associated with the low-frequency components, which are typically of the order of a rotor diameter, is not adequately captured by classical RDT approaches



170

175

180



(Mann et al., 2018; Ghate et al., 2018; Milne and Graham, 2019). Other theoretical studies based on linearized or quasi-steady fluctuation models, which show qualitative agreement with high-fidelity simulations and measurements, found that low-frequency eddies undergo significant distortion, whereas high-frequency ones are largely unaffected (Graham, 2017; Mann et al., 2018; Ghate et al., 2018; Milne and Graham, 2019). In terms of noise emissions, it should be noted that the sound generated by distorted large-scale turbulent structures in the inflow lies outside the audible range (Buck et al., 2018). IT noise for WTs, usually prevailing in the low-frequency range of sound waves, is then generated by the interaction of mid-to-small turbulent structures with blades. Yet, it remains unclear whether turbulence within the IT noise frequency band is exposed to noticeable distortion due to streamtube expansion and, if so, how to best model it. This highlights a key gap in current understanding.

This study examines the impact of turbulence distortion on IT noise in next-generation large-scale onshore WTs, considering the thickening blade profiles (Veers et al., 2019; Timmer and Bak, 2013; Schaffarczyk et al., 2024; Global, 2024) and intensifying inflow conditions that cannot be readily reproduced in wind tunnel experiments (Saab et al., 2018). This analysis is carried out using a developed Amiet's model framework. The distortion of turbulence is considered as two subsequent mechanisms. First, the high-frequency distortion due to streamtube expansion is evaluated using atmospheric flow simulations performed with the SOWFA (Simulator fOr Wind Farm Applications) (Fleming et al., 2013). Then, the distortion due to thick blade profiles is computed through a simplified closed-form solution of Goldstein's RDT associated only with vorticity deflections, allowing Amiet's model to maintain its low computational cost while enhancing its accuracy. The paper is structured as follows: after an overview of the Amiet-based IT noise tool and its adaptation to rotating frames, the turbulence distortion mechanisms and their implementations are presented. Finally, using the developed prediction model, the relevance of IT noise for future onshore WTs is analyzed, taking into account the effects of turbulence distortion.

2 IT noise prediction tool

The IT noise prediction tool used in this study is based on Amiet's model (Amiet, 1975), an analytical theory developed to estimate the far-field noise spectrum radiated from an infinite-span flat plate subjected to a subsonic turbulent gust. In the context of IT noise, the pressure jump across the flat plate that is responsible for acoustic radiation is considered to arise from the interaction of the LE with the incoming turbulence. The original configuration simply corresponds to an isolated airfoil in an acoustic wind tunnel where the observer and the airfoil are fixed relative to a reference frame, as illustrated in Fig. 1.

To extend this model to WT applications, the developed tool adapts Amiet's formulation to a WT rotating blade frame, by following well-established procedures for helicopters and rotors (Amiet, 1989; Sinayoko et al., 2013). This adaptation ensures that the fundamental assumptions of the model are still satisfied while accounting for the unsteady kinematics and spatial inhomogeneity present in actual WT operations.



205



2.1 Adaptation of Amiet's model to a WT reference frame

195 Consider an observer positioned far away from a WT, as shown in Fig. 2(a), with the reference frame initially located at the rotor hub. The adaptation procedure for this configuration consists of the following steps:

2.1.1 Blade segmentation

Each WT blade is discretized into multiple segments along the span to capture the variations in inflow turbulence and blade geometry (e.g., chord, twist, and airfoil profile). Each segment is modeled as an independent IT noise source under rectilinear motion, following the original Amiet's configuration (Fig. 1). A representative segment is highlighted in orange in Fig. 2(a).

2.1.2 Reference frame transformation

In order to apply Amiet's formulation, the observer distance and the incoming Mach number are expressed in the local frame of each segment. This new reference frame, (x_a, y_a, z_a) , is centered at the midspan of a blade segment and oriented according to Amiet's coordinate system (see Fig. 2(a)). Following the procedures described by Amiet (1989) and Sinayoko et al. (2013), the required coordinate transformation is carried out in three steps:

- The hub-based reference frame, (x_h, y_h, z_h) , is translated to the LE of the segment's midspan, resulting in (x_1, y_1, z_1) ;
- A rotation is applied about the z_1 -axis by the local azimuth angle, γ , aligning the translated frame with the segment span, as shown in Fig. 2(b);
- The resulting frame, (x_2, y_2, z_2) , is rotated about the y_2 -axis by an angle β , which is the local pitch angle (sum of the blade pitch and local twist angles) to ensure chord alignment, as shown in Fig. 2(c).

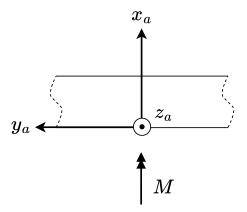
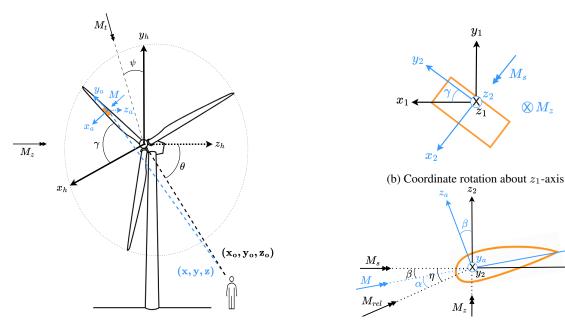


Figure 1. Original Amiet's coordinate system.







(a) Coordinate transformation from hub to Amiet's frame for a blade segment

(c) Coordinate rotation about y_2 -axis

Figure 2. Sketches depicting coordinate transformation for adapting Amiet's model to WT rotating frames.

This transformation yields the observer location, x, relative to the local Amiet-aligned source frame, (x_a, y_a, z_a) . The final coordinates are calculated using the rotation matrices as follows:

$$\mathbf{x} = R_y(\beta)R_z(\pi/2 - \gamma)(\mathbf{x}_o - \mathbf{x}_s)
= \begin{bmatrix} \cos(\beta) & 0 & \sin(\beta) \\ 0 & 1 & 0 \\ -\sin(\beta) & 0 & \cos(\beta) \end{bmatrix} \begin{bmatrix} \cos(\pi/2 - \gamma) & -\sin(\pi/2 - \gamma) & 0 \\ \sin(\pi/2 - \gamma) & \cos(\pi/2 - \gamma) & 0 \\ 0 & 0 & 1 \end{bmatrix} (\mathbf{x}_o - \mathbf{x}_s), \tag{1}$$

where x_o and x_s denote the observer and the present segment/source positions based on the hub, respectively. Using the angles for the same segment demonstrated in Fig. 2, the incoming Mach number, M, aligned with the x-axis as in the original configuration is obtained by

$$M = M_{rel}cos(\alpha)$$
. (2)

Here, M_{rel} is the relative Mach number between the incoming flow and the moving segment, and α is the local angle of attack, obtained as

$$M_{rel} = \sqrt{M_s^2 + M_z^2}$$

$$= \sqrt{(\Omega r + M_t cos(\psi + \gamma))^2 + M_z^2};$$
(3)





$$\alpha = \tan^{-1} \left(\frac{M_z}{M_s} \right) - \beta. \tag{4}$$

In this context, M_z and M_t are the axial and transversal components of the freestream Mach number, respectively, as seen in Fig. 2(a), while ψ defines the orientation of M_t relative to the y_h -axis. The azimuthal Mach number, M_s , is derived from the vector sum of the rotational blade Mach, $M_b = \Omega r/c_\infty$ and M_t , where Ω is the blade angular velocity, r is the radial location of the segment, and c_∞ is the speed of sound.

Lastly, the present source position, x_s , at the moment the emitted sound reaches the observer (i.e., the emission moment), is calculated by convecting the emission position forward within the acoustic propagation time. The approach, valid even under non-zero angle of attack conditions, follows the method described by Sinayoko et al. (2013) as

$$\mathbf{x}_{s} = r_{e} \{ -(M_{b}sin(\gamma) + M_{t}cos(\gamma)sin(\gamma + \psi))\hat{i} + (M_{b}cos(\gamma) - M_{t}sin(\gamma)sin(\gamma + \psi))\hat{j} \}.$$
(5)

Here, r_e denotes the distance between the observer and the retarded source positions, which is computed by

$$r_e = r_o \frac{M_c cos(\Theta) + \sqrt{1 - M_c^2 sin^2(\Theta)}}{1 - M_c^2},\tag{6}$$

where r_o is the direct observer distance from the hub, while M_c is the convection Mach number, which is simply equal to $\sqrt{M_z^2 + M_t^2}$. The angle Θ between M_c and \boldsymbol{x}_o is calculated as

235
$$\Theta = \cos^{-1}\left(\frac{M_t sin(\psi) sin(\theta) + M_z cos(\theta)}{M_s}\right),$$
 (7)

with θ being the angle of r_o relative to the z_h -axis.

2.1.3 Doppler shift effect

Fixing the reference frame to each segment makes the source be under rectilinear motion while the observer moves relatively, necessitating a Doppler shift correction to properly account for the circular motion in reality. Amiet's model is, thus, applied at the emission frequency, ω_e , whereas the observer perceives the sound at the Doppler-shifted frequency, ω . The spectral relationship at the observer position x is given by

$$S_{pp}(\boldsymbol{x},\omega,\gamma) = \frac{\omega_e}{\omega} S'_{pp}(\boldsymbol{x},\omega_e,\gamma), \tag{8}$$

where S'_{pp} is the radiated power spectral density (PSD) from a segment at the azimuthal angle of γ . The Doppler shift factor, ω_e/ω , is given as

245
$$\frac{\omega_e}{\omega} = \frac{1 + (\boldsymbol{M}_c - \boldsymbol{M}_b) \cdot \hat{s}}{1 + \boldsymbol{M}_c \cdot \hat{s}}.$$
 (9)

Here, the convection and blade Mach numbers are expressed in vector form as

$$\mathbf{M}_{c} = -M_{t} \sin(\psi)\hat{i} - M_{t} \cos(\psi)\hat{j} - M_{z}\hat{k},$$

$$\mathbf{M}_{b} = -M_{b} \sin(\gamma)\hat{i} + M_{b} \cos(\gamma)\hat{j}.$$
(10)





In addition, the unit direction vector from the retarded source to the observer, \hat{s} , is obtained as

$$\hat{s} = -(M_t \sin(\psi) + x_o/r_e)\hat{i} - (M_t \cos(\psi) + y_o/r_e)\hat{j} - (M_z + z_o/r_e)\hat{k}. \tag{11}$$

250 2.1.4 Application of Amiet's model

Amiet's model is applied to each segment by taking into account the inverse strip theory (Santana et al., 2016), which overcomes the infinite-span assumption of the model. This theory basically applies Amiet's formulation to two high-aspect-ratio segments separated by the actual span length of the corresponding segment, and then subtracts the resulting spectra. This ensures that the upwash spectrum, ϕ_{22} , and aeroacoustic transfer (airfoil response) function, \mathcal{L} , are nearly independent of the spanwise wavenumber (Graham, 1970), simplifying the spectral formulation (Eq. (20) in the work of Amiet (1975)). Accordingly, for the computation of the inflow turbulence sound spectrum radiated from the midspan of each segment, the following formulation is used:

$$S'_{pp}(\boldsymbol{x},\omega_e,\gamma) = \left(\frac{\rho_{\infty}\omega_e z c M}{2\sigma^2}\right)^2 \frac{b}{2} |\mathcal{L}|^2 S_{22}(\omega_e) l_y(\omega_e), \tag{12}$$

where ρ_{∞} denotes the freestream density, and c and b are the blade segment's chord and span, respectively. The term σ^2 is computed as $\sigma^2 = x^2 + (1 - M^2)(y^2 + z^2)$. The transfer function \mathcal{L} includes both LE scattering and TE back-scattering components, following the formulations provided by Amiet (1989) and Santana et al. (2016) (not given here for the sake of brevity). The PSD of the upwash velocity fluctuations, S_{22} , and the spanwise correlation length, l_y , are revealed with the infinite-span assumption, making the spanwise direction homogeneous. Hence, S_{22} is obtained from the 1-D turbulence energy spectrum, $\phi_{22}(k_x)$, as

265
$$S_{22}(\omega_e) = \frac{\phi_{22}(k_x)}{U},$$
 (13)

where U_x is the incoming streamwise velocity (i.e., in the x-axis), whereas k_x is the chordwise wavenumber that equals to $k_x = \omega_e/U_x$. The spanwise correlation length l_y is computed as

$$l_y(\omega_e) = \frac{15.02L\hat{k}_x^2}{(3+8\hat{k}_x^2)\sqrt{1+\hat{k}_x^2}},\tag{14}$$

with $\hat{k}_x = k_x/k_e$ where $k_e = 0.7468/L$, and L being the incoming turbulent length scale.

In the cases of isotropic incoming turbulence (i.e., neglecting distortion effects), $\phi_{22}(k_x)$ can be modeled by a von Kármán (vK) 1-D spectrum as follows:

$$\phi_{22}(k_x) = 0.0792 \frac{\overline{u_x^2}}{k_e} \frac{3 + 8\hat{k}_x^2}{\left(1 + \hat{k}_x^2\right)^{11/6}},\tag{15}$$

where u_x is the root-mean-square of fluctuating streamwise velocity. For other cases, the calculation of the distorted spectra is given in Sect. 4.1.





Table 1. Operational parameters of benchmark cases

Case no	$U_{\infty} [\mathrm{m s}^{-1}]$	Ω [RPM]	β_0 [deg]
1	6	13	3
2	8	14	-2
3	9.5	17	5

75 2.1.5 Azimuthal averaging

The segment-level noise predictions are azimuthally averaged over a full blade revolution to obtain the time-averaged spectrum. This introduces an additional Doppler factor, ω_e/ω , due to the rotational retarded time effect (Amiet, 1989; Sinayoko et al., 2013) (i.e., the change of time increment from source to observer). Eventually, the total IT noise radiated by the WT is obtained by summing the contributions from all segments of blades as follows:

280
$$S_{pp}(\boldsymbol{x},\omega) = \frac{B}{2\pi} \sum_{1}^{N_{\text{seg}}} \int_{0}^{2\pi} \left(\frac{\omega_e}{\omega}\right)^2 S'_{pp}(\boldsymbol{x},\omega_e,\gamma) d\gamma$$
 (16)

where B is the number of blades, and $N_{\rm seg}$ is the number of segments per blade.

2.2 Verification of IT noise tool

285

290

295

The implementation of the IT noise prediction tool described in Sect. 2.1 is verified using a benchmark case of an onshore WT for which far-field noise measurements are openly available from the zEPHYR EU-project (Christophe et al., 2022). The corresponding WT is the SIEMENS SWT-2.3-92 model, which features an $80\,\mathrm{m}$ high tower and a $92\,\mathrm{m}$ rotor diameter. As reported by Leloudas (2006), measured sound pressure levels at $100\,\mathrm{m}$ downstream of the turbine tower were converted into sound power level spectra, L_w , by accounting for background noise effects and other relevant measurement corrections. These spectra were provided for three operational cases varying with average wind speed U_∞ , rotational speed Ω , and blade pitch angle β_0 . The associated parameters are summarized in Table 1. The IT characteristics, including the turbulent length scale ($L=312\,\mathrm{m}$) and turbulence intensity (TI = 10.7%), were obtained from atmospheric boundary layer simulations performed by Kale (2024). In particular, these parameters were calculated for Case 2 and assumed unchanged for the other cases. In the present verification, turbulence distortion effects are not included; instead, the upwash velocity spectrum is calculated using the vK isotropic model (see Eq. (15)).

The IT noise spectra are predicted for all three operating conditions using the developed tool. Figure 3 shows a comparison between the predictions and the measured data. For additional verification, results from another independently developed Amiet-based IT noise tool (Botero-Bolívar, 2023; Botero-Bolívar et al., 2024) are also included in the comparison. Although this tool differs from the current implementation in terms of vK spectrum formulation, coordinate transformation, and applied strip theory, the agreement between the two prediction tools is nearly perfect across all frequencies and cases. This close match serves as a verification of the implementation of the model. On the other hand, both prediction tools demonstrate a fair



310



agreement with the measured data, particularly in the low-frequency range, as shown in Figs. 3(d-f), where IT noise is most pronounced. It is important to note that the TE noise is not modeled in either of the Amiet-based predictions, which explains the considerable underprediction at higher frequencies.

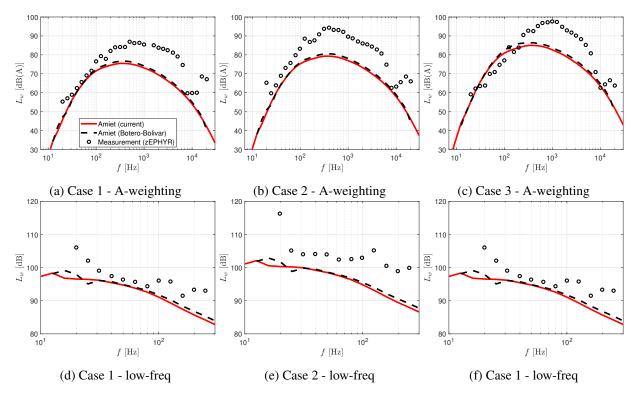


Figure 3. 1/3 octave sound power levels compared with results from another Amiet tool (Botero-Bolívar, 2023) and measured data (Christophe et al., 2022).

3 Turbulence distortion due to streamtube expansion

This section investigates whether the incoming turbulence undergoes distortion during the streamtube expansion within the rotor's induction zone, with a particular focus on the frequency range relevant for IT noise. To this end, atmospheric flow simulations are performed around a large WT using the SOWFA open-source tool (Fleming et al., 2013), which basically couples the solvers OpenFOAM and OpenFAST. In this study, the flow field is simulated by Large-Eddy Simulation (LES) via OpenFOAM while modeling the rotor blade forces by Actuator Line Model via OpenFAST and projecting them to the LES grid simultaneously. The coupled framework offers a cost-effective yet sufficiently accurate solution, especially considering that accurate resolution of unsteady blade aerodynamics and wake structures is not a primary concern here. All simulations are carried out on the Delft High Performance Computing Centre (DHPC).



320

325

330



3.1 Simulation setup of SOWFA

The simulations are conducted using the NREL 5-MW reference WT, which is widely studied in the literature and whose configuration files are publicly available. Two different inflow wind speeds are considered: a below-rated case at $8\,\mathrm{m\,s^{-1}}$ and an above-rated case at $14\,\mathrm{m\,s^{-1}}$. These two regimes allow for an examination of the potential influence of the axial induction factor on turbulence distortion.

Each case starts with a precursor simulation, in which the atmospheric flow is simulated without any turbine. Here, the goal is to establish a turbulent flow that has a neutral atmospheric boundary layer with a desired wind speed at the hub height of 90 m. The computational domain size is $3 \,\mathrm{km} \times 3 \,\mathrm{km} \times 1 \,\mathrm{km}$ in the streamwise (x), spanwise (y), and surface-normal (z) directions, respectively. All horizontal boundary conditions are treated by periodicity. The streamwise and spanwise grid resolution is $10 \,\mathrm{m}$, whereas the grid is clustered toward the surface. The time step is chosen as $0.5 \,\mathrm{s}$. The precursor simulations are run for $20000 \,\mathrm{s}$ to obtain statistically-converged turbulence.

For the below-rated case, a streamwise velocity spectrum of a random point at hub height is presented in Fig. 4. The spectrum is compared with the vK spectrum, which is obtained based on the sampled turbulence intensity and length scale. The results show that the flow at the hub height exhibits an isotropic turbulence behavior as expected, up to a numerical cut-off frequency imposed by the solver capabilities with the current grid and timestep.

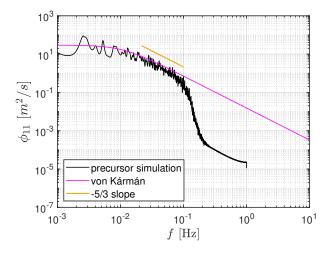


Figure 4. Streamwise velocity spectrum of a random point at hub height.

3.2 Energy spectrum change in induction zone

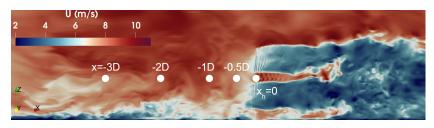
The last $2000 \, \mathrm{s}$ of the flow data from the precursor simulation are stored and then used to initialize the main turbine-included simulation, serving as the initial flow domain as well as the inflow boundary condition. The grid is refined around the turbine rotor in 4 levels, with the smallest cell sizes becoming $0.625 \, \mathrm{m}$. Accordingly, the time step is also reduced to $0.125 \, \mathrm{s}$. The finest grid domain extends up to 3 rotor diameters (D) upstream of the rotor plane in order to analyze turbulence distortion in the induction zone without grid effects. Besides, the streamwise periodic boundaries are replaced by inflow and outflow conditions.



350



Each simulation with the modified setup and configuration is continued for 2000 s. The turbine rotational speed is fixed at 9 RPM for both wind speed cases. During the simulations, the velocity time histories are sampled at various points in the induction zone for a period of nearly 225 revolutions of a blade at a sampling frequency of 80 Hz. The instantaneous side and cross-sectional views of streamwise velocity contours, along with the sampled data locations at hub height, are shown in Fig. 5. The turbine and its wake within the atmospheric boundary layer are clearly visible.





(a) Side view (x-z plane)

(b) Cross view (y-z plane)

Figure 5. Streamwise velocity contours from different views, together with some sampled data locations.

The streamwise velocity component is selected for spectral analysis, as it directly corresponds to the upwash velocity with respect to the blade, which is the one responsible for inducing the unsteady surface pressure fluctuations and hence IT noise generation. The streamwise evolution of the mean and fluctuating components of this velocity in the induction zone is plotted in Fig. 6 for both inflow cases and three different spanwise locations. Theoretical data based on the actuator disk model and the quasi-steady formulations (Mann et al., 2018) are also included for comparison. Accordingly, the theoretical distortions of the mean (*U*) and fluctuating (*u*) speed components as the flow approaches the rotor are defined as

$$\frac{U(x)}{U_{\infty}} = 1 - a_x \left(1 + \frac{\xi}{\sqrt{1 + \xi^2}} \right)$$

$$\frac{u^2(x)}{u_{\infty}^2} = 1 - a_x \left(1 + \frac{\xi}{\sqrt{1 + \xi^2}} \right) \left(1 + \frac{U_{\infty}}{a_x} \frac{da_x}{dU_{\infty}} \right),$$
(17)

where a_x is the axial induction factor, and $\xi=2x/D$. The induction factor and its derivative with respect to the inflow speed are obtained from steady-state OpenFAST simulations over a range of wind speeds. Figure 6 shows that the simulated mean velocities match the theoretical predictions relatively well, while fluctuating components show only qualitative agreement, consistent with observations in earlier studies (Mann et al., 2018; Milne and Graham, 2019), which questions the validity of the linearized actuator disk theory close to the rotor. This observed decrease of the streamwise turbulence intensity in the below-rated case could be attributed to a rotor-induced blockage effect, as explained by the physical findings of the RDT. In contrast, the above-rated case shows an amplification, with the vortex line pile-up effect likely becoming the dominant mechanism (Hunt, 1973). This difference can be related to variations in the induction factor and its gradient, both of which modulate turbulence distortion via changes in aerodynamic loading. This is also consistent with the observed reduction in distortion towards the blade tip, where the loading is minimal.



360

365



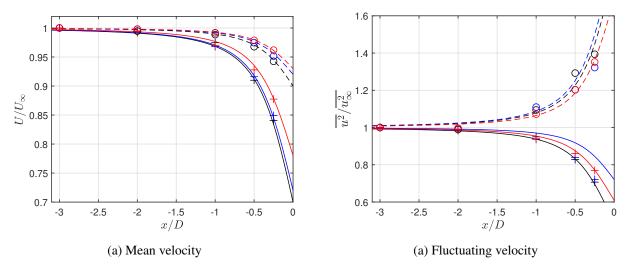


Figure 6. Change of mean and fluctuating streamwise velocity along the induction zone. Sampled data locations are shown in Figure 5. The symbols + and o denote the simulation data, whereas solid and dashed lines are generated by the theoretical formula for below- and above-rated speeds, respectively. Black, blue and red colors refer to the spanwise locations of y_h , $y_h + 0.25D$ and $y_h + 0.40D$, respectively.

These averaged quantities primarily reflect the dynamics of large-scale eddies, the sizes of which are larger than the rotor diameter in both wind speed cases. Such large-scale structures correspond to very low frequencies that contribute little to audible noise radiation. Therefore, the analysis is continued by focusing on the streamwise velocity spectrum at mid-to-high frequencies, evaluated at various upstream positions for the below-rated case. A representative comparison at a spanwise location of 0.4D, where the IT noise contribution is expected to be strong, is depicted in Fig. 7. The first observation is that the spectrum reveals the harmonics of the blade passing frequency $(0.45\,\mathrm{Hz})$ near the rotor, superimposed on a broadband turbulence background. Additionally, when compared to the vK model, each simulated spectrum exhibits a numerical cut-off beyond which the inertial subrange slope (-5/3) is reduced. This shift in cut-off frequency on the finer grid suggests that the observed change is numerical in origin, rather than a physical distortion. The same conclusion is reached for the above-rated speed case (not shown here). These results must be evaluated considering the capacity of the SOWFA tool, where the near field of the rotor is modeled. Nevertheless, the findings are consistent with those of linearized-theory-based studies (Mann et al., 2018; Ghate et al., 2018; Milne and Graham, 2019), where the distortion of small-scale eddies was reported as negligible. Thus, based on these investigations, turbulence distortion due to the streamtube expansion is neglected in the remaining part of this study for the frequency range of interest.





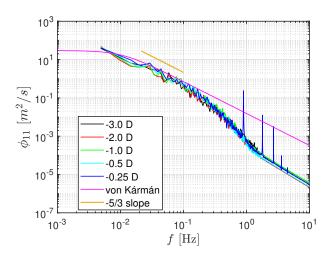


Figure 7. Change of streamwise energy spectrum along the upstream rotor axis $(y_h + 0.4D, z_h)$.

4 Turbulence distortion due to blade thickness

Following the streamtube expansion, inflow turbulence is additionally affected by the blade thickness as it approaches the blade LEs. In this section, this mechanism is examined using the RDT of Goldstein (1978). A practical implementation of this approach for aeroacoustic applications has already been presented by Glegg and Devenport (2017), including a systematic method to compute distorted spectra from undistorted ones. Their formulation relies on a distortion tensor approximated for high-frequency gusts by Majumdar and Peake (1998). In this study, following the steps of Glegg and Devenport (2017), a generalized version of this approximated distortion tensor is derived, which remains valid across all frequency ranges, not limited to the high-frequency regime. The goal of this derivation is to increase the accuracy in Amiet-based IT noise predictions for WT blades with large distortion-inducing body thicknesses. The following section presents how to obtain this tensor starting from the fundamental assumptions and initial steps of Goldstein's RDT. In the end, its capability is discussed within the concept of classical distortion mechanisms.

4.1 Goldstein's RDT

380

The RDT framework assumes the following conditions:

- The incoming turbulence is weak: $u_{\infty} \ll U_{\infty}$.
- The distorted unsteady flow is isentropic and incompressible.
- The mean flow is irrotational and homentropic, implying that boundary layer effects and flow separation are neglected.

¹The mean flow does not have to be strictly irrotational; it is the irrotational component that primarily governs turbulence distortion.





- The distortion time scale is much shorter than the turbulent time scale. This is the main principle of RDT, meaning the disturbances are convected only by the mean flow.

Goldstein's RDT approach starts with decomposing the unsteady velocity as

$$v = U + \nabla \varphi + u^{(g)}. \tag{18}$$

where v is the mean velocity, φ is the velocity potential term related to pressure fluctuations, and $u^{(g)}$ is the gust velocity including the remaining potential and rotational disturbances. At the upstream boundary, the unsteady flow is determined by the gust velocity while the velocity potential being zero. Under the above assumptions, the linearized Euler equations reduce to the wave equation for the velocity potential as

$$\frac{1}{c_{\infty}^2} \frac{D_0^2 \varphi}{Dt^2} - \nabla^2 \varphi = \nabla \cdot \boldsymbol{u}^{(g)},\tag{19}$$

where D_0/Dt is the substantial derivative relative to the mean flow. As the disturbances are convected by an irrotational mean flow, the upstream gust velocity can be tracked by the drift coordinates analytically, which also allows Eq. (19) to be uncoupled. Drift coordinates (X_1, X_2, X_3) were initially introduced by Darwin (1953) and then Lighthill (1956) as

$$\frac{D_0 X_1}{Dt} = U_{\infty}, \quad \frac{D_0 X_2}{Dt} = 0, \quad \frac{D_0 X_3}{Dt} = 0.$$
 (20)

Here, the constant X_1 identifies the surface representing the fluid locations after the same particles are convected with the mean flow velocity for a certain time (X_1/U_∞) , whereas X_2 and X_3 are the stream surfaces orthogonal to each other (not to X_1).

400 X_1 is obtained via

$$X_1 = U_{\infty} \int_{\Gamma} \frac{d\sigma_s}{U},\tag{21}$$

with σ_s denoting the distance from the upstream boundary along the streamline. After obtaining all drift coordinates and their gradients (numerically or analytically), one can obtain the gust velocity in regions where the RDT assumptions are valid as follows:

405
$$\boldsymbol{u}^{(g)} = \nabla X_j u_j^{(\infty)}(X_2, X_3, t - X_1/U_\infty),$$
 (22)

where $u_i^{(\infty)}$ refers to the undistorted velocity fluctuation at the upstream boundary. Given the upstream fluctuating velocity as a summation of harmonic waves as

$$\boldsymbol{u}^{(\infty)} = \sum_{k} Re\left(\hat{\boldsymbol{u}}^{(\infty)}e^{-i\omega t}\right) = \sum_{k} Re\left(\hat{\boldsymbol{a}}e^{i\boldsymbol{k}\cdot\boldsymbol{x}}e^{-i\omega t}\right),\tag{23}$$

the distorted gust velocity can also be written in the wavenumber domain as follows:

410
$$\mathbf{u}^{(g)} = \sum_{k} Re\left(\hat{\mathbf{u}}^{(g)}e^{-i\omega t}\right) = \sum_{k} Re\left(\nabla X_{j}\hat{a}_{j}e^{i\mathbf{k}\cdot\mathbf{X}}e^{-i\omega t}\right),$$
 (24)





where \hat{a}_i is the complex amplitude of the gust velocity at each wavenumber vector k_i , and $\omega = k_1 U_{\infty}$. Note that, at the upstream boundary, x and X are equal to each other.

The next step is to introduce the velocity potential, φ , which represents the irrotational part of the disturbance field. In practice, φ can be expressed in terms of harmonic components, which allows the governing wave equation to be solved for the particular solution associated with the gust velocity². Accordingly, if φ is defined as

$$\varphi = \sum_{k} Re\left(\hat{\varphi}e^{-i\omega t}\right) = \sum_{k} Re\left(\hat{Q}e^{i\mathbf{k}\cdot\mathbf{X}}e^{-i\omega t}\right),\tag{25}$$

the wave equation given in Eq. (19) becomes

$$\nabla^2 \hat{\varphi} = -\nabla \cdot \hat{\boldsymbol{u}}^{(g)}. \tag{26}$$

Here, the Laplace and the divergence operators can be found as follows:

$$\begin{split} \nabla \hat{\varphi} &= i \nabla (\boldsymbol{k} \cdot \boldsymbol{X}) \hat{Q} e^{i \boldsymbol{k} \cdot \boldsymbol{X}} \\ &= i k_j \nabla X_j \hat{Q} e^{i \boldsymbol{k} \cdot \boldsymbol{X}} \\ &= i \boldsymbol{\kappa} \hat{Q} e^{i \boldsymbol{k} \cdot \boldsymbol{X}}, \\ \nabla^2 \hat{\varphi} &= (i k_j \nabla^2 X_j + i \boldsymbol{\kappa} i \boldsymbol{\kappa}) \hat{Q} e^{i \boldsymbol{k} \cdot \boldsymbol{X}} \\ &= (i k_j \nabla^2 X_j - |\boldsymbol{\kappa}|^2) \hat{Q} e^{i \boldsymbol{k} \cdot \boldsymbol{X}}, \end{split}$$

420
$$\nabla \cdot \hat{\boldsymbol{u}}^{(g)} = (i\boldsymbol{\kappa} \cdot \nabla X_k + \nabla^2 X_k) \hat{a}_k e^{i\boldsymbol{k} \cdot \boldsymbol{X}},$$
 (27)

where $\kappa = \nabla X_j k_j$, representing the distorted wavenumber domain. Eventually, the magnitude of the velocity potential term, \hat{Q} , is obtained by solving Eq. (26) as

$$\begin{split} \nabla^2 \hat{\varphi} &= -\nabla \cdot \hat{\boldsymbol{u}}^{(g)} \\ (ik_j \nabla^2 X_j - |\boldsymbol{\kappa}|^2) \hat{Q} e^{i\boldsymbol{k} \cdot \boldsymbol{X}} &= -(i\boldsymbol{\kappa} \cdot \nabla X_k + \nabla^2 X_k) \hat{a}_k e^{i\boldsymbol{k} \cdot \boldsymbol{X}}; \end{split}$$

425

$$\hat{Q} = \frac{(i\boldsymbol{\kappa} \cdot \nabla X_k + \nabla^2 X_k)\hat{a}_k}{|\boldsymbol{\kappa}|^2 - ik_j \nabla^2 X_j}.$$
(28)

Under the high-frequency gust assumption mentioned above, the expression is simplified by neglecting high-order terms such that

$$\hat{Q} \approx \frac{i\boldsymbol{\kappa} \cdot \nabla X_k \hat{a}_k}{|\boldsymbol{\kappa}|^2}.$$
(29)

²In the present formulation, only this particular solution is retained, while the homogeneous correction that satisfies the exact surface boundary condition is not considered.





430 Then, the amplitude of the distorted velocity can be found as

$$\hat{\boldsymbol{u}} = \nabla \hat{\varphi} + \hat{\boldsymbol{u}}^{(g)}$$

$$= \left(i \kappa \frac{i \kappa \cdot \nabla X_k}{|\kappa|^2} + \nabla X_k \right) \hat{a}_k e^{i \boldsymbol{k} \cdot \boldsymbol{X}}$$

$$= \left(-\kappa \frac{\kappa \cdot \nabla X_k}{|\kappa|^2} + \nabla X_k \right) \hat{u}_k^{(\infty)},$$
(30)

which can also be expressed in terms of a distortion tensor, D_{ij} , as

435
$$\hat{u}_i(\kappa) = D_{ij}\hat{u}_i^{(\infty)}(\mathbf{k}),$$
 (31)

where

$$D_{ij} = \frac{\partial X_j}{\partial x_i} - \frac{\kappa_i \kappa_k}{|\kappa|^2} \frac{\partial X_j}{\partial x_k}.$$
(32)

For use in spectral models, as in Amiet's formulation, it is better to compute the distorted energy spectrum directly, which was provided by Glegg and Devenport (2017) through the same distortion tensor as

$$440 \quad \phi_{ij}(\boldsymbol{\kappa}) = D_{ip}D_{iq}\phi_{pq}^{(\infty)}(\boldsymbol{k}), \tag{33}$$

where ϕ_{ij} represents the 3-D energy spectrum. Then, the 1-D spectrum is obtained by integrating over k_2 and k_3 , to be used in the current Amiet's tool.

In this study, to remove the high-frequency assumption and account for all eddy sizes, a more general form of the distortion tensor is derived by directly substituting Eq. (28) in Eq. (30) as follows:

$$\hat{\boldsymbol{u}} = \nabla \hat{\varphi} + \hat{\boldsymbol{u}}^{(g)}$$

$$= \left(i\kappa \frac{(i\kappa \cdot \nabla X_k + \nabla^2 X_k)}{|\kappa|^2 - ik_j \nabla^2 X_j} + \nabla X_k\right) \hat{a}_k e^{i\mathbf{k} \cdot \mathbf{X}}$$

$$= \left(\frac{-|\kappa|^2 \kappa (\kappa \cdot \nabla X_k) - k_j \nabla^2 X_j \kappa \nabla^2 X_k + i \left(|\kappa|^2 \kappa \nabla^2 X_k - k_j \nabla^2 X_j \kappa (\kappa \cdot \nabla X_k)\right)}{|\kappa|^4 + (k_j \nabla^2 X_j)^2} + \nabla X_k\right) \hat{a}_k e^{i\mathbf{k} \cdot \mathbf{X}}.$$
(34)

Then, the new distortion tensor becomes

$$D_{ij} = \frac{\partial X_j}{\partial x_i} - \frac{1}{|\boldsymbol{\kappa}|^4 + (k_k \nabla^2 X_k)^2} \left\{ |\boldsymbol{\kappa}|^2 \kappa_i \kappa_k \frac{\partial X_j}{\partial x_k} + k_k \nabla^2 X_k \kappa_i \frac{\partial^2 X_j}{\partial x_k^2} - i \left(|\boldsymbol{\kappa}|^2 \kappa_i \frac{\partial^2 X_j}{\partial x_k^2} - k_k \nabla^2 X_k \kappa_i \kappa_k \frac{\partial X_j}{\partial x_k} \right) \right\}.$$
(35)

While more complex in form, this derived tensor remains computationally efficient, as all required derivatives are precomputed.

It should be mentioned that Goldstein's RDT, in the present form, does not incorporate the blockage mechanism. This limitation arises because only the scattering potential is solved, and the exact surface boundary condition is not enforced. In this way, a practical, closed-form solution of the RDT is obtained. This omission mainly affects the lowest frequencies, which are outside the range of interest for the present WT applications. Indeed, the present form of Goldstein's RDT includes the vorticity



460



deflection mechanism of the classical RDT approach. Here, the physical interpretation of the high-frequency approximation is that the size of the disturbances is assumed to be smaller than the distortion length scale, $l_{\rm dist}$, which is on the order of the characteristic length of the body. Mathematically, this implies $kl_{\rm dist}\gg 1$, where k is the corresponding gust wavenumber. However, in WT applications, atmospheric eddies often have length scales larger than blade chords, challenging the high-frequency gust restriction. This motivates the use of the generalized version of the tensor (derived without any frequency restriction) in Sect. 5, where IT noise generation from WTs is predicted by Amiet's model based on RDT. In view of the above, given an upstream undistorted energy spectrum, the distorted spectrum can be computed using Eqs. (33) and (35), within the region where the RDT assumptions are valid.

4.2 Verification of the generalized distortion tensor

The derived distortion tensor is first verified in a wind tunnel problem where the incoming incompressible flow is axisymmetrically contracting or expanding. In this case, the gradient tensor of the drift coordinates is known as

$$\frac{\partial X_j}{\partial x_i} = \begin{pmatrix} 1/AR & 0 & 0\\ 0 & \sqrt{AR} & 0\\ 0 & 0 & \sqrt{AR} \end{pmatrix},\tag{36}$$

where AR is the area ratio between the inlet and the outlet. Here, AR < 1 refers to contraction (mean flow acceleration), whereas AR > 1 means expansion (mean flow deceleration).

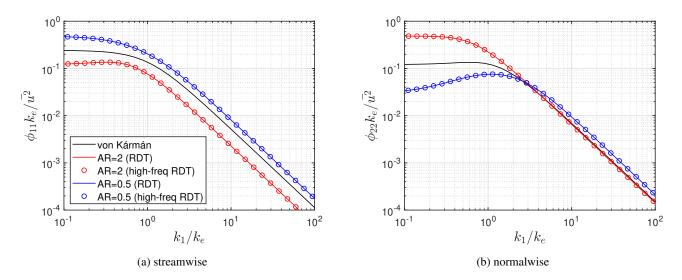


Figure 8. 1-D energy spectra of turbulence distorted by axisymmetric wind tunnel contraction (AR = 0.5) and expansion (AR = 2).

1-D spectra of streamwise and normalwise velocity fluctuations distorted in a wind tunnel are computed by the RDT for both contraction and expansion cases, and they are shown in Fig. 8. For comparison, the same spectra obtained by using the high-frequency approximated distortion tensor (hereafter, called as high-freq RDT) by Glegg and Devenport (2017) are included





470 as well. The results match exactly, as expected, since all second derivatives of drift coordinates are zero, yielding that both distortion tensors become identical to each other.

The derived distortion tensor is further tested in a problem of a 2-D incompressible turbulent flow approaching a circular cylinder, where experimental data close to the cylinder LE is available (Britter et al., 1979). This benchmark case is particularly important because the distortion mechanism of a WT blade section is modeled with a representative circular cylinder in this study, which is described in Sect. 5.

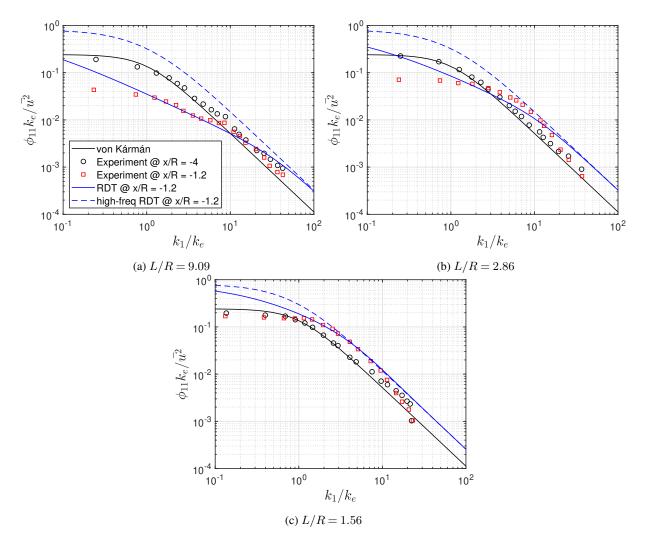


Figure 9. 1-D streamwise energy spectra of turbulence distorted by a circular cylinder. Experimental data is obtained from Britter et al. (1979).

The incoming flow has a turbulent intensity of 5% approximately. The RDT computations are conducted at an upstream location of x/R = -1.2 for three different turbulent length scale (L) to cylinder radius (R) ratios (L/R = [9.09, 2.86, 1.56]).



480

485

490

500

505

510



The 3-D vK model is used to estimate the undistorted energy spectra, which allow the computation of the 1-D distorted ones via the RDT. Figure 9 compares measured and predicted streamwise spectra, along with results from the high-freq RDT. The experimental data available at x/R = -4.0 and the 1-D vK isotropic spectra are also plotted as references to the undistorted turbulence.

The results show that the vK model accurately matches the undistorted flow at x/R = -4.0, providing that the flow is out of distortion effects due to the presence of the cylinder at that position. At x/R = -1.2, both RDT formulations capture high-frequency distortion in small eddies $(k_1/k_e \gg 1)$ due to vorticity deflection, following the typical RDT mechanisms in the asymptotic limits. However, the RDT with the generalized form of the high-frequency distortion tensor (hereafter, simply denoted by RDT) shows better agreement in the mid-frequency range $(1 < k_1/k_e < 10)$, particularly for larger eddies. Notably, the high-freq RDT, in contrast to the RDT, does not vary with L/R and fails to capture the changes through lower frequencies.

The superiority of the generalized version arises because it depends explicitly on the magnitude of wavenumber in the distortion tensor, making the turbulence–geometry interaction sensitive to the turbulence length scale relative to the body size. In this way, the distortion becomes sensitive to the characteristic dimension of the eddies compared to the obstacle radius. In contrast, the high-frequency approximation treats all modes as short waves, and thereby filtering out large eddies (low wavenumbers). This suppresses any dependence on L and leads to inaccurate mid-frequency predictions. In the context of WTs, IT noise generally corresponds to a wider frequency range $(k_1/k_e > 1)$, justifying the use of the generalized version in subsequent predictions.

495 5 Analysis of RDT-based IT noise predictions: An evaluation of future WTs

The RDT is integrated into the Amiet-based IT noise prediction tool (called Amiet-RDT hereafter) by replacing the upwash energy spectrum in Eq. (12) with its distorted counterpart. The RDT computations are performed at locations near the stagnation point of each blade profile. The distortion mechanism is applied by assuming that the incoming turbulence approaches a representative circle whose radius is the distortion length of the corresponding airfoil. This distortion length is estimated by the length between the LE radius and half of the maximum thickness of the airfoil. This approach is based on the findings of a recent study (Piccolo et al., 2024), which showed that an airfoil distorts turbulence similarly to a cylinder with a radius equal to its distortion length. That study proposed using the arc length between the stagnation point and the point of maximum curvature as a measure of the distortion length. This approach avoids the need to recompute the drift coordinates for each airfoil profile, making the RDT implementation computationally efficient.

The developed Amiet-RDT tool is applied to the benchmark WT configuration used in Sect. 2.2 to assess how turbulence distortion affects the IT noise spectrum. Case 1 is used for this analysis, and the results are compared with those obtained by the Amiet tool, which is based on the vK isotropic model. The upwash spectra in the vicinity of the LEs are shown in Fig. 10 (a) for different blade sections, the span location of which is specified by z/R, R being the rotor radius. The corresponding distortion lengths ($l_{\rm dist}$) are also given. From root to tip, $l_{\rm dist}$ decreases, as expected from typical blade profile thicknesses. The results indicate that, within the low-frequency range of interest for IT noise, the distorted spectra are not significantly different from the





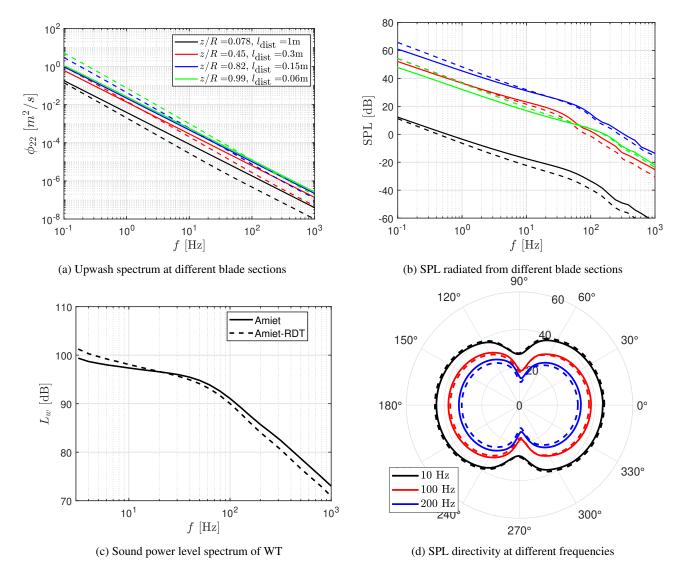


Figure 10. Impact of RDT-based distortion on IT noise prediction for the benchmark WT configuration used in Sect. 2.2. Dashed lines refer to the RDT calculations.

isotropic ones. This is because the relevant frequencies fall within the transition region between the two RDT asymptotic limits. As a result, the sound pressure level (SPL) contribution from each section remains almost unchanged, as shown in Fig. 10 (b). An exception is observed close to the blade root, which has a circular profile and the largest $l_{\rm dist}$. A larger $l_{\rm dist}$ increases the range of high-frequency eddies affected by distortion, leading to more small-scale structures with reduced energy. However, these inner blade regions contribute minimally to overall noise, so the total generated IT noise (quantified in terms of sound power level, L_w , in Fig. 10 (c)) is predicted nearly identically from both Amiet and Amiet-RDT models. The typical IT noise



520

540

545

550



directivity pattern on the horizontal ground plane, shown in Fig. 10 (d), where 0 degree corresponds to the observer location, confirms this result as well.

To assess the potential impact of turbulence distortion in next-generation, larger WTs, the same analysis is repeated for a conceptual WT scaled to reflect the anticipated growth in onshore turbine sizes (Veers et al., 2019). In this configuration, the rotor radius of the benchmark turbine is tripled, and the blade chords are scaled linearly with the span. Although industry practice suggests that chord growth tends to be more than linear near the root and less than linear near the tip (Bortolotti et al., 2016, 2019), this parametric study focuses on relative changes in turbulence distortion with increasing thickness. Hence, a uniform geometric scaling of the chord provides a plausible first-order assumption (Canet et al., 2021).

For the blade thicknesses, however, a more than linear scaling is applied. For very large turbines, designers tend to increase relative thickness to maintain strength and stiffness without excessively increasing the blade mass (Timmer and Bak, 2013; Caboni et al., 2017). In the benchmark case, airfoils at mid-to-outer span have a LE radius of about 0.02c and a maximum thickness of about 0.18c, where c refers to the chord. In the next-generation turbines, LE radii of approximately 0.05c and maximum thicknesses of 0.35–0.40c are expected, according to industry indications and literature (Timmer and Bak, 2013; Schaffarczyk et al., 2024). Accordingly, the ratio of distortion length to chord of each section, except near the root where the circular profile exists, is doubled in the conceptual configuration. Finally, the same flow and operating conditions as Case 1 are retained, except that the RPM is reduced by one-third to preserve comparable relative velocities with respect to the benchmark case.

Figure 11 demonstrates the influence of turbulence distortion on noise predictions for this enlarged turbine, as previously. Corresponding increases in $l_{\rm dist}$ for the same sections are shown in Fig. 11 (a). The main finding is that thicker blade profiles, under the same inflow conditions, cause a significant distortion of the turbulent velocity field, which is observed by the alteration of the corresponding spectrum within the frequency range of interest (see Fig. 11 (a)). Across all blade sections, the distortion lies in the high-frequency regime of RDT, where vorticity line deflections reduce the energy of smaller eddies. As a result, thicker profiles yield reduced IT noise, a trend clearly visible in both the overall sound power levels and directivity patterns, as shown in Figs. 11(c-d). This reduction increases with frequency. When compared with the benchmark configuration in Fig. 11 (c), the standard Amiet model predicts a monotonic rise in noise with rotor size, whereas the Amiet-RDT tool shows that spectral alterations mitigate this increase. At higher frequencies, the larger turbine can even produce noise levels comparable to the benchmark configuration.

The above analysis isolates thickness effects while keeping the inflow and operating conditions the same. Figures 12(a-b) demonstrate the same analysis by varying the incoming turbulence length scale for the conceptual rotor. For this case, the wind speed and the turbine rotational speed are set to $6\,\mathrm{m\,s^{-1}}$ and 13 RPM, respectively. At higher frequencies, corresponding to smaller eddies, noise levels scale inversely with the length scale since this quantity mainly reflects the energetic large eddies, as also reported by Botero-Bolívar et al. (2024). Within this regime, the Amiet-RDT model consistently predicts lower noise than the baseline Amiet approach. Changes in length scale also shift the spectral shape, which is further influenced directly by turbulence distortion. As a result, the Amiet-RDT model can even predict higher low-frequency noise, as observed for $L=10\,\mathrm{m}$.





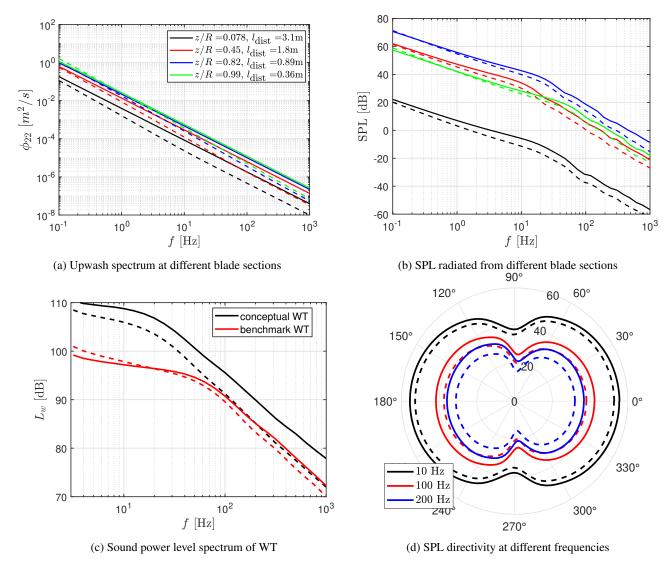


Figure 11. Impact of RDT-based distortion on IT noise prediction for the conceptual WT configuration. Dashed lines refer to the RDT calculations.

Another scenario is studied by fixing the length scale to $L=10\,\mathrm{m}$ and varying the rotational speed between 1 and 16 RPM. The results are shown in Figs. 12(c-d). Within the IT noise frequency range, the predicted noise levels increase proportionally to the rotational speed, in contrast to the previous case. Again, due to the change in spectral shape induced by the distortion effect, the differences between the Amiet and Amiet–RDT predictions depend on both frequency and speed. In particular, the distortion effect caused by the thicker WT profiles is found to be more evident at lower rotational speeds.





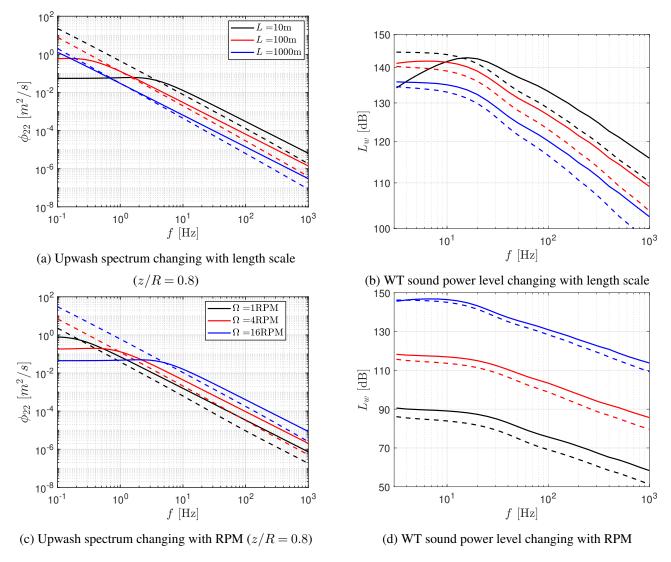


Figure 12. Assessment of IT noise predictions with and without RDT for different turbulent length scales (a-b) and different turbine rotational speeds (c-d). Dashed lines refer to the RDT calculations.

These results imply that IT noise levels do not necessarily scale proportionally with rotor size once realistic blade thickening is accounted for in future WTs. Turbulence distortion changes the balance across frequencies, reducing high-frequency noise but sometimes amplifying low-frequency contributions. Consequently, trends predicted with the canonical spectra or the empirical models based on certain frequency ranges may not remain valid for future large-scale designs.



565

570

575

580

585

590



6 Conclusions

This study investigates the relevance of IT noise for next-generation large WTs by explicitly accounting for turbulence distortion effects considering their growing rotor sizes, an aspect overlooked in existing low-fidelity approaches relying on empirical corrections. In this sense, an accurate, computationally efficient, and physics-based IT noise prediction tool based on Amiet's analytical model is developed. Two main mechanisms are examined: streamwise spectral changes in the induction zone and distortion near thick blade LEs.

The turbulence distortion due to streamtube expansion in the induction zone is examined by LES coupled with modeling the blade forces via SOWFA. The spectral results reveal that turbulence distortion in the induction region remains negligible in the mid-to-high frequency range, which is relevant for acoustic scattering. This result suggests that the upstream turbulence spectra estimated by isotropic models can be used for this region without reducing accuracy.

To address distortion due to blade thickness that appears as the upstream turbulence approaches the blade surfaces, a practical form of Goldstein's RDT, which is associated with vorticity deflections, is developed. Unlike earlier approaches, this model does not rely on the high-frequency gust approximation and remains valid across a broad frequency range within its capabilities. It is implemented into an Amiet-based IT noise tool through a distortion tensor formulation, which directly modifies the upwash spectrum near the blade LEs. The method is validated against canonical test cases, and its accuracy in capturing the vorticity deflection effects of the RDT mechanism is demonstrated.

Importantly, the developed Amiet-RDT tool modifies the input turbulence spectrum based on geometry (via a representative distortion length), operational conditions, and inflow properties. This eliminates the need for empirical coefficients commonly used to adjust isotropic energy spectrum models, which are not suitable for realistic scales of atmospheric turbulence and WTs. This tool increases prediction accuracy without additional computational cost, making the developed model practical for parametric noise assessments. It is believed that this model will be especially beneficial in the design and development of WT noise mitigation strategies.

When applied to a conceptual onshore WT scaled through an existing benchmark configuration, the developed tool reveals that turbulence distortion near blade surfaces becomes increasingly influential for future turbines. This is particularly the case as relative thicknesses are expected to scale more than linearly - reaching around 0.4c from the current $\sim 0.2c$ - to maintain structural strength without excessive mass growth. The results show that IT noise levels exhibit nonlinear behavior with rotor size, contrary to what is predicted by standard models. The distortion effects, particularly the suppression of small-scale eddy energy by vorticity deflection, can even reduce overall noise generation at certain frequencies for a larger rotor where higher noise would normally be expected based on geometric scaling alone. The results also indicate that turbulence distortion interacts with the inflow length scale and rotational speed, leading to frequency-dependent shifts in IT noise levels. This reinforces that scaling trends in the noise spectrum cannot be generalized without accounting for both geometric and operational factors.

As a recommended next step, a high-fidelity simulation of the entire WT flow field, including near-blade domain, can be performed. This approach would allow a detailed investigation of near-blade turbulence distortion and provide valuable insight



600



regarding possible rotor-induced distortion effects at wider frequency ranges. Such work could serve to further validate the RDT-based predictions.

Data availability. The data that support the findings of this study are available from the corresponding author, ÖY, upon reasonable request.

Author contributions. All authors contributed to the conceptualization of the study. ÖY and AP developed the Amiet-based prediction tool. ÖY carried out the SOWFA simulations and subsequent postprocessing. ÖY, AP, and RZ contributed to the investigations and derivations of the RDT formulation. ÖY performed the data analysis, visualization, and validation of the results. DR and RM acquired the funding and coordinated the project. ÖY wrote the manuscript, and all authors contributed to its reviewing and editing.

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

Acknowledgements. This paper is part of the Blade Extensions for Silent Turbines (BEST) project (HER+22-02-03415120) which is supported by the Netherlands Enterprise Agency (Rijksdienst voor Ondernemend Nederland (RVO)), acting on behalf of the Dutch Ministry of Economic Affairs, using the Renewable Energy Transition subsidy (subsidie Hernieuwbare Energietransitie (HER+ 2022-02)), part of the Topsector Energy (TSE) subsidies.

The authors gratefully acknowledge Dr. Dachuan Feng and Nirav Dangi for their support with running the SOWFA simulations on the DelftBlue cluster.





References

- Amiet, R.: Acoustic radiation from an airfoil in a turbulent stream, Journal of Sound and Vibration, 41, 407–420, https://doi.org/10.1016/s0022-460x(75)80105-2, 1975.
 - Amiet, R. K.: Noise produced by turbulent flow into a rotor: theory manual for noise calculation, Tech. rep., NASA Langley Research Center, 1989.
 - Atassi, H., Subramaniam, S., and Scott, J.: Acoustic radiation from lifting airfoils in compressible subsonic flow, in: 13th Aeroacoustics Conference, American Institute of Aeronautics and Astronautics, https://doi.org/10.2514/6.1990-3911, 1990.
- Batchelor, G. K. and Proudman, I.: THE EFFECT OF RAPID DISTORTION OF A FLUID IN TURBULENT MOTION, The Quarterly Journal of Mechanics and Applied Mathematics, 7, 83–103, https://doi.org/10.1093/qjmam/7.1.83, 1954.
 - Blumendeller, E., Kimmig, I., Huber, G., Rettler, P., and Cheng, P. W.: Investigations on Low Frequency Noises of On-Shore Wind Turbines, Acoustics, 2, 343–365, https://doi.org/10.3390/acoustics2020020, 2020.
- Bortolotti, P., Bottasso, C. L., and Croce, A.: Combined preliminary–detailed design of wind turbines, Wind Energy Science, 1, 71–88, https://doi.org/10.5194/wes-1-71-2016, 2016.
 - Bortolotti, P., Tarrés, H., Dykes, K., Merz, K., Sethuraman, L., Verelst, D., and Zahle, F.: IEA Wind TCP Task 37: Systems Engineering in Wind Energy-WP2.1 Reference Wind Turbines, National Renewable Energy Laboratory (NREL), https://doi.org/10.2172/1529216, 2019. Botero-Bolívar, L.: Wind Turbine Noise Prediction, https://doi.org/10.5281/zenodo.8325973, 2023.
- Botero-Bolívar, L., Marino, O. A., Venner, C. H., de Santana, L. D., and Ferrer, E.: Low-cost wind turbine aeroacoustic predictions using actuator lines, Renewable Energy, 227, 120 476, https://doi.org/10.1016/j.renene.2024.120476, 2024.
 - Britter, R. E., Hunt, J. C. R., and Mumford, J. C.: The distortion of turbulence by a circular cylinder, Journal of Fluid Mechanics, 92, 269–301, https://doi.org/10.1017/s0022112079000628, 1979.
 - Buck, S., Oerlemans, S., and Palo, S.: Experimental characterization of turbulent inflow noise on a full-scale wind turbine, Journal of Sound and Vibration, 385, 219–238, https://doi.org/10.1016/j.jsv.2016.09.010, 2016.
- Buck, S., Oerlemans, S., and Palo, S.: Experimental Validation of a Wind Turbine Turbulent Inflow Noise Prediction Code, AIAA Journal, 56, 1495–1506, https://doi.org/10.2514/1.j056134, 2018.
 - Caboni, M., Boorsma, K., and Kanev, S.: Development of thick airfoils for outboard sections and investigation into their application for large rotors, Wind Engineering, 42, 177–193, https://doi.org/10.1177/0309524x17736480, 2017.
- Canet, H., Bortolotti, P., and Bottasso, C. L.: On the scaling of wind turbine rotors, Wind Energy Science, 6, 601–626, https://doi.org/10.5194/wes-6-601-2021, 2021.
 - Christophe, J.: Application of hybrid methods to high frequency aeroacoustics, Phd thesis, von Karman Institute for Fluid Dynamics, 2011.
 - Christophe, J., Buckingham, S., Schram, C., and Oerlemans, S.: zEPHYR Large On Shore Wind Turbine Benchmark, https://doi.org/10.5281/zenodo.6380878, 2022.
- Darwin, C.: Note on hydrodynamics, in: Mathematical Proceedings of the Cambridge Philosophical Society, vol. 49, pp. 342–354, Cambridge University Press, 1953.
 - Davy, J. L., Burgemeister, K., and Hillman, D.: Wind turbine sound limits: Current status and recommendations based on mitigating noise annoyance, Applied Acoustics, 140, 288–295, https://doi.org/10.1016/j.apacoust.2018.06.009, 2018.
 - Delft High Performance Computing Centre (DHPC): DelftBlue Supercomputer (Phase 2), https://www.tudelft.nl/dhpc/ark:/44463/DelftBluePhase2, 2024.





- Devenport, W. J., Staubs, J. K., and Glegg, S. A.: Sound radiation from real airfoils in turbulence, Journal of Sound and Vibration, 329, 3470–3483, https://doi.org/10.1016/j.jsv.2010.02.022, 2010.
 - dos Santos, F. L., Botero-Bolívar, L., Venner, C. H., and de Santana, L. D.: Inflow turbulence distortion for airfoil leading-edge noise prediction for large turbulence length scales for zero-mean loading, The Journal of the Acoustical Society of America, 153, 1811–1822, https://doi.org/10.1121/10.0017458, 2023.
- Dos Santos, F. L., Botero-Bolívar, L., Vennner, K., and De Santana, L. D.: A new approach to account for turbulence distortion effect on Amiet's leading-edge noise prediction, INTER-NOISE and NOISE-CON Congress and Conference Proceedings, 270, 4340–4351, https://doi.org/10.3397/in_2024_3447, 2024.
 - Faria, A. M., Saab, J. Y., Rodriguez, S., and de Mattos Pimenta, M.: A rapid distortion theory-based airfoil turbulent inflow noise prediction method, Journal of the Brazilian Society of Mechanical Sciences and Engineering, 42, https://doi.org/10.1007/s40430-020-02468-2, 2020.
- Fleming, P., Gebraad, P., van Wingerden, J. W., Lee, S., Churchfield, M., Scholbrock, A., Michalakes, J., Johnson, K., and Moriarty, P.: SOWFA Super-Controller: A High-Fidelity Tool for Evaluating Wind Plant Control Approaches, National Renewable Energy Lab. (NREL), Golden, CO (United States), https://doi.org/10.2172/1090162, 2013.
 - Ghate, A. S., Ghaisas, N., Lele, S. K., and Towne, A.: Interaction of small scale Homogenenous Isotropic Turbulence with an Actuator Disk, in: 2018 Wind Energy Symposium, American Institute of Aeronautics and Astronautics, https://doi.org/10.2514/6.2018-0753, 2018.
- Glegg, S. and Devenport, W.: Aeroacoustics of low Mach number flows: fundamentals, analysis, and measurement, Academic Press, ISBN 978-0-12-809651-2, https://doi.org/10.1016/B978-0-12-809651-2.09991-4, 2017.
 - Global, S.: SANY Renewable Energy Successfully Installs the 15-MW Onshore Wind Turbine, https://www.sanyglobal.com/press_releases/3922/#:~:text=On%20October%208%2C%20a%2015,for%20long%2Ddistance%20land%20transportation, Accessed: 9 October 2025, 2024
- Global Wind Energy Council: Global Wind Report 2024, https://www.gwec.net/reports/globalwindreport/2024, Accessed: 9 October 2025, April 2024.
 - Goldstein, M. E.: Unsteady vortical and entropic distortions of potential flows round arbitrary obstacles, Journal of Fluid Mechanics, 89, 433–468, https://doi.org/10.1017/s0022112078002682, 1978.
- Graham, J. M. R.: Similarity rules for thin aerofoils in non-stationary subsonic flows, Journal of Fluid Mechanics, 43, 753–766, https://doi.org/10.1017/s0022112070002719, 1970.
 - Graham, J. M. R.: Rapid distortion of turbulence into an open turbine rotor, Journal of Fluid Mechanics, 825, 764–794, https://doi.org/10.1017/jfm.2017.400, 2017.
 - Hunt, J. C. R.: A theory of turbulent flow round two-dimensional bluff bodies, Journal of Fluid Mechanics, 61, 625–706, https://doi.org/10.1017/s0022112073000893, 1973.
- Kale, B.: ABL simulations with uncertain weather parameters and impact on WT performance and near-field noise, Ph.D. thesis, Universidad Politécnica de Madrid University Library, https://doi.org/10.20868/upm.thesis.80430, 2024.
 - Kirkegaard, J. K., Cronin, T. H., Nyborg, S., and Frantzen, D. N.: The multiple understandings of wind turbine noise: reviewing scientific attempts at handling uncertainty, Wind Energy Science, 10, 907–924, https://doi.org/10.5194/wes-10-907-2025, 2025.
- Lahoz, M., Nabhani, A., Saemian, M., and Bergada, J. M.: Wind Turbine Enhancement via Active Flow Control Implementation, Applied Sciences, 14, 11 404, https://doi.org/10.3390/app142311404, 2024.
 - Leloudas, G.: Optimization of Wind Turbines with Respect to Noise, Master's thesis, Technical University of Denmark, MSc thesis. Available at: https://outer.space.dtu.dk/~giorgos/images/giorgosleloudas2006.pdf, 2006.





- Lighthill, M. J.: Drift, Journal of Fluid Mechanics, 1, 31–53, https://doi.org/10.1017/s0022112056000032, 1956.
- Lowson, M. V.: Assessment and prediction of wind turbine noise, Tech. rep., Flow Solutions Ltd., Bristol (UK), 1993.
- Majumdar, S. J. and Peake, N.: Noise generation by the interaction between ingested turbulence and a rotating fan, Journal of Fluid Mechanics, 359, 181–216, https://doi.org/10.1017/s0022112097008318, 1998.
 - Mann, J., Peña, A., Troldborg, N., and Andersen, S. J.: How does turbulence change approaching a rotor?, Wind Energy Science, 3, 293–300, https://doi.org/10.5194/wes-3-293-2018, 2018.
- Milne, I. A. and Graham, J. M. R.: Turbulence velocity spectra and intensities in the inflow of a turbine rotor, Journal of Fluid Mechanics, 870, https://doi.org/10.1017/jfm.2019.339, 2019.
 - Mish, P. F. and Devenport, W. J.: An experimental investigation of unsteady surface pressure on an airfoil in turbulence—Part 1: Effects of mean loading, Journal of Sound and Vibration, 296, 417–446, https://doi.org/10.1016/j.jsv.2005.08.008, 2006.
 - Moreau, S. and Roger, M.: Effect of Angle of Attack and Airfoil Shape on Turbulence-Interaction Noise, in: 11th AIAA/CEAS Aeroacoustics Conference, American Institute of Aeronautics and Astronautics, https://doi.org/10.2514/6.2005-2973, 2005.
- Moriarty, P. and Migliore, P.: Semi-Empirical Aeroacoustic Noise Prediction Code for Wind Turbines, Tech. rep., National Renewable Energy Laboratory, https://doi.org/10.2172/15006098, 2003.
 - Moriarty, P., Guidati, G., and Migliore, P.: Prediction of Turbulent Inflow and Trailing-Edge Noise for Wind Turbines, in: 11th AIAA/CEAS Aeroacoustics Conference, American Institute of Aeronautics and Astronautics, https://doi.org/10.2514/6.2005-2881, 2005.
- Møller, H. and Pedersen, C. S.: Low-frequency noise from large wind turbines, The Journal of the Acoustical Society of America, 129, 3727–3744, https://doi.org/10.1121/1.3543957, 2011.
 - Oerlemans, S., Sijtsma, P., and Méndez López, B.: Location and quantification of noise sources on a wind turbine, Journal of Sound and Vibration, 299, 869–883, https://doi.org/10.1016/j.jsv.2006.07.032, 2007.
 - Paterson, R. and Amiet, R.: Acoustic radiation and surface pressure characteristics of an airfoil due to incident turbulence, in: 3rd Aeroacoustics Conference, American Institute of Aeronautics and Astronautics, https://doi.org/10.2514/6.1976-571, 1976.
- Piccolo, A., Zamponi, R., Avallone, F., and Ragni, D.: Turbulence-distortion analysis for leading-edge noise-prediction enhancement, in: AIAA AVIATION 2023 Forum, American Institute of Aeronautics and Astronautics, https://doi.org/10.2514/6.2023-3628, 2023.
 - Piccolo, A., Zamponi, R., Avallone, F., and Ragni, D.: Turbulence distortion and leading-edge noise, Physics of Fluids, 36, https://doi.org/10.1063/5.0244627, 2024.
- Piccolo, A., Zamponi, R., Avallone, F., and Ragni, D.: Modification of Amiet's model for turbulence-ingestion noise prediction in rotors, The

 Journal of the Acoustical Society of America, 158, 461–475, https://doi.org/10.1121/10.0037185, 2025.
 - Rogers, A. L. and Manwell, J. F.: Wind turbine noise issues, Tech. rep., Renewable Energy Research Laboratory, University of Massachusetts, Amherst, 2004.
 - Saab, J. Y., de Mattos Pimenta, M., Faria, A. M., and Rodriguez, S.: Determination of local flow conditions and preliminary investigation on the validity of the PNoise code for large-size wind turbine, Journal of the Brazilian Society of Mechanical Sciences and Engineering, 40, https://doi.org/10.1007/s40430-018-1412-1, 2018.
 - Santana, L. D., Christophe, J., Schram, C., and Desmet, W.: A Rapid Distortion Theory modified turbulence spectra for semi-analytical airfoil noise prediction, Journal of Sound and Vibration, 383, 349–363, https://doi.org/10.1016/j.jsv.2016.07.026, 2016.
 - Schaffarczyk, A. P., Lobo, B. A., Balaresque, N., Kremer, V., Suhr, J., and Wang, Z.: Development and Measurement of a Very Thick Aerodynamic Profile for Wind Turbine Blades, Wind, 4, 190–207, https://doi.org/10.3390/wind4030010, 2024.





- 720 Sinayoko, S., Kingan, M., and Agarwal, A.: Trailing edge noise theory for rotating blades in uniform flow, Proceedings of the Royal Society A: Mathematical, Physical and Engineering Sciences, 469, 20130 065, https://doi.org/10.1098/rspa.2013.0065, 2013.
 - Tian, Y., Cotté, B., and Chaigne, A.: Wind Turbine Noise Modelling Based on Amiet's Theory, in: Proceedings of the 5th International Meeting on Wind Turbine Noise, pp. CD–ROM proceedings, Denver, CO, United States, https://ensta.hal.science/hal-00975237, 2013.
- Timmer, W. and Bak, C.: Aerodynamic characteristics of wind turbine blade airfoils, in: Advances in Wind Turbine Blade Design and Materials, Woodhead Publishing Series in Energy, pp. 109–149, Woodhead Publishing, ISBN 978-0-85709-426-1, https://doi.org/https://doi.org/10.1533/9780857097286.1.109, 2013.
 - Veers, P., Dykes, K., Lantz, E., Barth, S., Bottasso, C. L., Carlson, O., Clifton, A., Green, J., Green, P., Holttinen, H., et al.: Grand challenges in the science of wind energy, Science, 366, eaau2027, https://doi.org/10.1126/science.aau2027, 2019.
- Wind Europe: Wind energy in Europe: 2024 Statistics and the outlook for 2025-2030, https://windeurope.org/data/products/wind-energy-in-europe-2024-statistics-and-the-outlook-for-2025-2030, Accessed: 9 October 2025, February 2025.
 - Zamponi, R., Satcunanathan, S., Moreau, S., Ragni, D., Meinke, M., Schröder, W., and Schram, C.: On the role of turbulence distortion on leading-edge noise reduction by means of porosity, Journal of Sound and Vibration, 485, 115 561, https://doi.org/10.1016/j.jsv.2020.115561, 2020.
- Zamponi, R., Moreau, S., and Schram, C.: Rapid distortion theory of turbulent flow around a porous cylinder, Journal of Fluid Mechanics, 915, https://doi.org/10.1017/jfm.2021.8, 2021.
 - Zhou, T., Cao, H., Zhang, M., and Liao, C.: Performance simulation of wind turbine with optimal designed trailing-edge serrations, Energy, 243, 122 998, https://doi.org/10.1016/j.energy.2021.122998, 2022.