



# A numerical study of the influence of terrain on wakes, blockage, wind farm efficiency, and turbine efficiency

James Bleeg<sup>1</sup>

<sup>1</sup>Group Research and Development, DNV, Bristol, BS1 6EW, UK

5 Correspondence to: James Bleeg (james.bleeg@dnv.com)

## Abstract.

This study investigates the interplay of terrain, blockage, and wake effects using Reynolds-Averaged Navier-Stokes (RANS) simulations of 35 different combinations of terrain, wind farm layout, and atmospheric conditions. The terrain includes two idealized solitary ridgelines, an idealized valley, and flat ground. The wind farms comprise one or two rows of closely spaced turbines parallel to the terrain feature. We simulate these idealized wind farms in conventionally neutral boundary layers of different heights. The set of simulations also includes an existing onshore wind farm located along a ridgeline and run with stable and unstable surface conditions. The horizontal variation of the ground elevation (i.e. terrain) has a large influence on wake and blockage effects in this study. In addition, the predicted wind farm efficiency and turbine efficiency (power coefficient) vary significantly depending upon the terrain in the simulation and the position of the wind farm relative to the terrain. For single-row wind farms the predicted effect of terrain on wind farm efficiency can exceed 4% – for the simulated conditions. The separate but correlated effect of terrain on individual turbine efficiency is of a similar magnitude. Analysis of the results indicates that there are multiple physical drivers behind the efficiency trends, including streamwise pressure gradients and inviscid effects related to buoyancy. Energy prediction methods that do not account for these drivers have an elevated risk of producing large errors – at least at wind farms similar to those evaluated in the study.

## 20 1 Introduction

An energy yield analysis for a planned onshore wind farm requires two types of flow simulations: freestream and wind farm. Freestream simulations include terrain but not, by definition, wind turbines; wind farm simulations include turbines but usually not terrain. The objective of a freestream flow analysis is to translate processed wind measurements, typically taken at just a few on-site locations, to the expected freestream wind conditions at each one of the planned turbine locations. Together with a turbine power curve, these wind conditions are used to estimate the energy yield for each turbine were it to operate in isolation. The sum of the standalone turbine energy yields is called the gross energy of the wind farm. A wind farm flow analysis calculates how wakes and blockage modify the wind conditions experienced by the turbines. More to the point, it quantifies for each turbine how much the energy yield changes when the turbine is operating in the wind farm compared with isolated operation. This change is known as a turbine interaction loss, which is usually the largest loss



30 contributing to the difference between the gross and net energy of a wind farm. While the effect of terrain on freestream (i.e. turbine-free) flow is a major consideration in energy yield assessments, the effect of terrain on wind farm flow – on wakes and on blockage – is usually neglected.

Several papers published over the last fifteen years report results indicating that hills can substantially affect wind turbine  
35 wake recovery. These include studies based on Reynolds-Averaged Navier-Stokes simulations (Politis et al., 2012), large-eddy simulations (Shamsoddin and Porté-Agel, 2018a; Liu and Stevens, 2019; Patel et al., 2024), and wind tunnel measurements (Tian et al., 2013; Hyvarinen et al. 2018). The Shamsoddin and Porté-Agel (2018) paper also validated a pressure-gradient-sensitive reduced-order wake model against the hill large-eddy simulations (LES). Based on the validation and a follow-on sensitivity study, the authors concluded that hill-induced streamwise pressure gradients can have a strong  
40 influence on wake recovery – with favorable pressure gradients on the windward side of the hill promoting wake recovery and adverse pressure gradients on the leeward side slowing wake recovery. (A pressure gradient is favorable when pressure decreases in the direction of flow; it is adverse when pressure increases in the direction of flow.) Dar and Porté-Agel (2022) further developed the reduced-order model and validated it against wind tunnel measurements of escarpment-affected wake flow. The results also pointed to the significant influence of terrain-induced pressure gradients on wake recovery.

45 Additional studies provide more direct evidence of streamwise pressure gradients influencing turbine wakes. Cai et al. (2021) varied the inclination of the floor of a wind tunnel to induce streamwise pressure gradients just downstream of a model wind turbine. Measured streamwise pressure variation was strongly correlated with and likely the cause of substantial differences in wake recovery. Favorable pressure gradients (FPGs) promoted wake recovery, and adverse pressure gradients (APGs) did the opposite. Bayron et al. (2024) induced streamwise pressure variation by adjusting the wind tunnel ceiling just downstream of the model turbine. The wake recovery was much slower in the APG wind tunnel configuration compared with the FPG configuration, with the zero pressure gradient (ZPG) wake recovery rate in between. In addition, the wake was vertically thicker and higher in the APG case compared with the ZPG case. The opposite trend was evident in the FPG case. Finally, Zhang et al. (2023) induced streamwise pressure gradients in LES by inclining the upper and lower walls to vary the  
55 cross-sectional areas of the domain along a row of four turbines aligned with the oncoming flow. They found that APGs made wakes deeper and wider, and FPGs did the opposite.

Collectively, these studies show that terrain-induced streamwise pressure gradients can substantially influence wake recovery. By implication, terrain can influence how much a turbine wake affects the wind speeds experienced by turbines  
60 located downstream, which affects wind farm efficiency. Other research suggests that the influence of terrain on the wake from an individual turbine also affects the wind conditions at the turbine itself. In other words, terrain can change turbine induction and thereby turbine power coefficient (i.e. individual turbine efficiency).



Before proceeding we offer a brief aside to define the blockage-related terms used in this paper. *Blockage* is an obstruction to the otherwise free flow of air. *Blockage effects* refer to the inviscid response of the flow to the obstruction. The terms “blockage” and “blockage effects” are often used interchangeably in the literature, and we will do the same here. *Wind farm blockage* refers to blockage from an array of wind turbines, and *turbine blockage* refers to blockage from an individual turbine. *Induction* is the turbine-blockage-related wind speed reduction upstream of the turbine, and *induction factor* is the fractional amount by which the wind speed at the rotor face of an isolated turbine is lower than the freestream wind speed.

In LES results reported in Trolborg et al. (2022), the power coefficient of a standalone turbine was found to be sensitive to terrain-induced streamwise gradients. The turbine in this study was at the top of a ridgeline, and different streamwise gradients were achieved by varying surface roughness, which affected the separation downstream of the ridge. In simulations where the hub-height freestream wind speed downstream of the turbine location increased with distance from the turbine, the efficiency of the turbine was higher. When the simulated freestream wind speed decreased with distance downstream of the turbine, the efficiency was lower. The findings were similar in Zengler et al. (2024), a sensitivity study run with a Reynolds-Averaged Navier-Stokes (RANS) model, where a single turbine was simulated atop various idealized hill geometries with different surface roughnesses. In both of these studies, streamwise variation in the freestream wind speed downstream of the turbine was found to correlate with changes in induction factor and in turn power coefficient. Dar et al. (2023) used wind tunnel experiments to investigate the effect of streamwise pressure gradients on wakes and power performance. Fifteen different tunnel floor configurations were tested. These configurations induced streamwise pressure gradients, progressing from a strong APG case to a ZPG case and through to a strong FPG case. Pressure was not directly measured in the experiments, rather it was inferred from the measured freestream wind speeds. In each configuration, the inferred freestream pressure gradient was nearly constant over a distance covering the induction zone of the model turbine through to the far wake. As the freestream streamwise pressure gradient increased, from the strong FPG case through the ZPG case and on to the strong APG case, the wake deficits generally also increased, while power coefficient generally decreased. For a given pressure gradient magnitude, an APG generally reduced turbine performance more than an FPG increased it. In an LES-based sensitivity study, Revaz and Porté-Agel (2024) simulated a standalone turbine atop various idealized hills. APGs downstream of the turbine increased wake deficits, which in turn increased turbine induction factor, decreasing power coefficient.

The impact of terrain on wind farm and turbine efficiency is still a relatively new field of study, with many opportunities to broaden the scope of the literature. Most of the research to date concerns wakes from a single turbine or a few aligned turbines. The results imply an effect of terrain on turbine interaction loss, but there is little in the literature directly making this connection through the analysis of full wind farms. In addition, the studies do not include the effect of buoyancy; the simulated and/or measured flow is always neutrally stratified. Yet wind farms are subjected to stably stratified flow all the time – if not within the boundary layer, then above it. This is significant because buoyancy effects profoundly modify the



response of flow to terrain (Baines, 2022; Vosper, 2004; Bleeg et al., 2015b; Radünz et al., 2025) and to wind farms (Lanzilao and Meyer, 2024). Finally, the peer-reviewed literature does not address the potential impact of terrain on wind farm blockage, which can have a material impact on the wind speeds experienced by turbines in an array (Bleeg et al., 2018).

This contribution takes aim at these gaps. Using a RANS solver, we simulate full wind farms in different types of idealized terrain. The simulated boundary layers are neutral but with overlying stratification. From these simulations we assess the influence of terrain on both wake *and* blockage effects. In turn, we evaluate the impact of terrain on turbine interaction loss across the wind farm. We also assess the impact of terrain on individual turbine efficiency. Finally, to better understand how the findings from these idealized simulations might translate to the field, we run additional simulations of a currently operating wind farm with inflow conditions derived from a numerical weather prediction model.

The next section of this paper, Section 2, explains the numerical approach used in this study. Section 3 presents the results, highlighting the effect of terrain on wakes, blockage, and energy conversion efficiency. The section also explores the main physical drivers behind the trends with terrain. Section 4 discusses the practical implications of the results and the limitations of the study. And lastly, Section 5 summarizes the main findings.

## 2 Numerical approach

This section describes the flow modeling used in the investigation. It describes the flow equations, boundary conditions, and domain, as well as how wind turbines are represented in the simulations. It also includes information about the simulated wind farms, atmospheric conditions, and terrain – and a description of the mesh used to resolve these features. Lastly, the section explains the types of simulations run and how they are post-processed to assess whether and to what degree turbine and wind farm performance are influenced by terrain.

There are two types of wind farms in this study: idealized and real. The numerical analysis of idealized wind farms – comprising 31 different combinations of turbine layout, terrain, and inflow conditions – represents the heart of the study. The one real wind farm in the investigation involves far fewer simulations, the results of which are included here as a check on the representativeness and real-world applicability of the idealized wind farms results. This section applies to both types of wind farms; however, the subsections covering terrain (2.2), wind farm layouts (2.3), and atmospheric conditions (2.4) apply primarily to the idealized wind farms. We address these three topics for the real wind farm in results section 3.4.

### 2.1 RANS model

We constructed and customized the numerical model within STAR-CCM+, a general-purpose computational fluid dynamics code (Siemens PLM, 2022). The mass and momentum equations are steady-state incompressible RANS equations. The



energy equation is a transport equation for potential temperature, and the turbulence equations are standard  $k-\epsilon$  with modified coefficients. Buoyancy is included in the vertical momentum equation via a shallow Boussinesq formulation; the effect of buoyancy on turbulence is modeled with production terms in the turbulence equations. The horizontal momentum equations include the Coriolis force, and the vertical momentum equation has a source term designed to damp gravity waves near the lateral and top boundaries. More details about the flow model may be found in Bleeg et al. (2015a and 2015b).

We represent the wind turbines with simple actuator disks. The thrust and torque, applied as body forces at each disk, are functions of the average axial velocity across the disk ( $U_{disk}$ ). The derivation of these functions is discussed further in Section 2.3.

This model for wind farm flows has been validated against field observations related to wake and/or blockage effects at 17 wind farms, including those publicly reported in Bleeg et al. (2018), Montavon et al. (2021), Montavon et al. (2023), Montavon et al. (2024), and Bleeg et al. (2024).

## 2.2 Terrain definition and domain

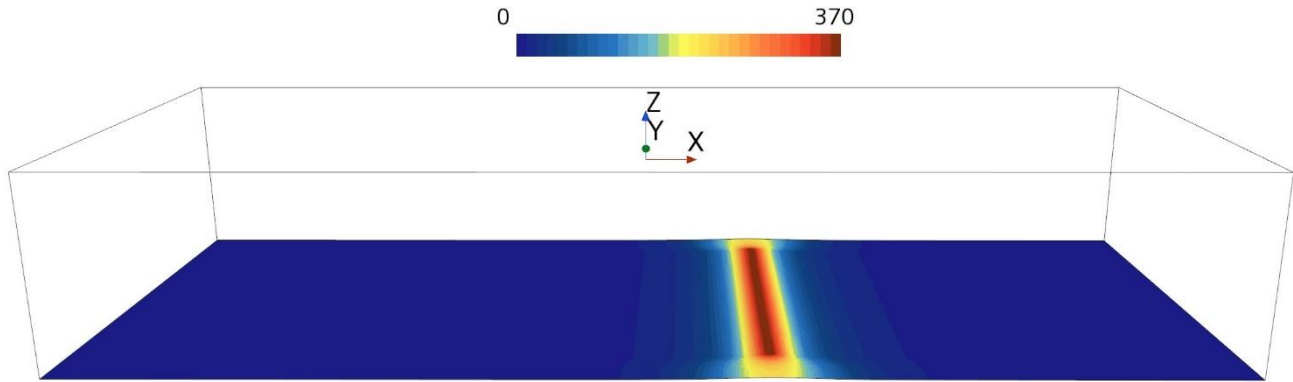
There are two primary terrain types in this study: flat and hill. In flat simulations, the ground elevation is uniformly equal to 0 m. In hill simulations, the ground elevation is a stretched Witch of Agnesi function (Eq. (1)) extruded north and south to create a quasi-two-dimensional hill in the three-dimensional RANS simulation. The height of the hill,  $h_c$ , is 370 m, and the base of the hill is at  $z = 0$  m. The half-height, half-width of the hill,  $a$ , is 2100 m in all hill cases but one, where  $a$  is set to 1050 m. The maximum slopes of these two hills are  $6.5^\circ$  and  $12.9^\circ$ , respectively. The flow does not separate from the ground surface in any of the simulations run with these hills. The study also includes a valley terrain type, with the valley being a mirror image of the quasi-2D hill ( $h_c = -370$  m).

$$z_{WOA} = \frac{h_c a^2}{a^2 + x^2} \quad (1)$$

The computational domain, depicted in Fig. 1, is large compared to the scales of the wind farm and hill profile. The length of the domain is 100 km with the western boundary at  $x = -60$  km and the eastern boundary at  $x = 40$  km; the center of the quasi-2D terrain feature is at  $x = 0$  km. The width of the domain is 40 km with the northern boundary at  $y = 20$  km and the southern boundary at  $y = -20$  km. The top boundary is at  $z = 17$  km. The wind direction at hub height is approximately  $270^\circ$ , and the geostrophic wind is from the northwest. Thus, the western and northern boundaries are inflow boundaries, where we specify velocity, turbulence quantities, and potential temperature. The southern and eastern boundaries are outflow



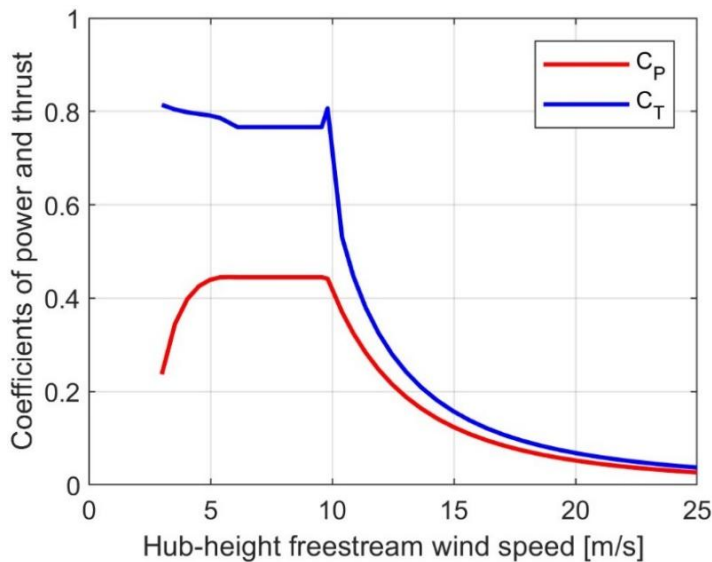
boundaries, where we specify static pressure. The top boundary is a slip wall set to a constant potential temperature. The  
 160 bottom, ground boundary is a no-slip wall with a uniform aerodynamic roughness length of 0.028 m and a constant potential  
 temperature of 283 K. The ground elevation is smoothed near the north and south boundaries, consistent with the standard  
 practice for simulating flow over terrain with this RANS model.



**Figure 1: Simulation domain with the ground surface colored by elevation (m).**

### 165 2.3 Wind farm and turbine definitions

The idealized wind farms are organized in rows. Each row has 34 turbines, and the hub-to-hub spacing between neighboring  
 turbines is two rotor diameters ( $2D$ ). The rows are north-to-south, perpendicular to the flow, and located either atop the hill  
 crest,  $15D$  downstream of the crest, or  $15D$  upstream of the crest. The turbine model has a hub height of 115 m, a rotor  
 diameter of 160 m, and a rated power of 5.1 MW. The  $C_T$  and  $C_P$  curves, depicted in Fig. 2, are from the IEA-3.4-130  
 170 onshore reference turbine (Bortolotti et al., 2019).



**Figure 2: Nominal power and thrust coefficient curves for the wind turbine model simulated in the idealized wind farms.**

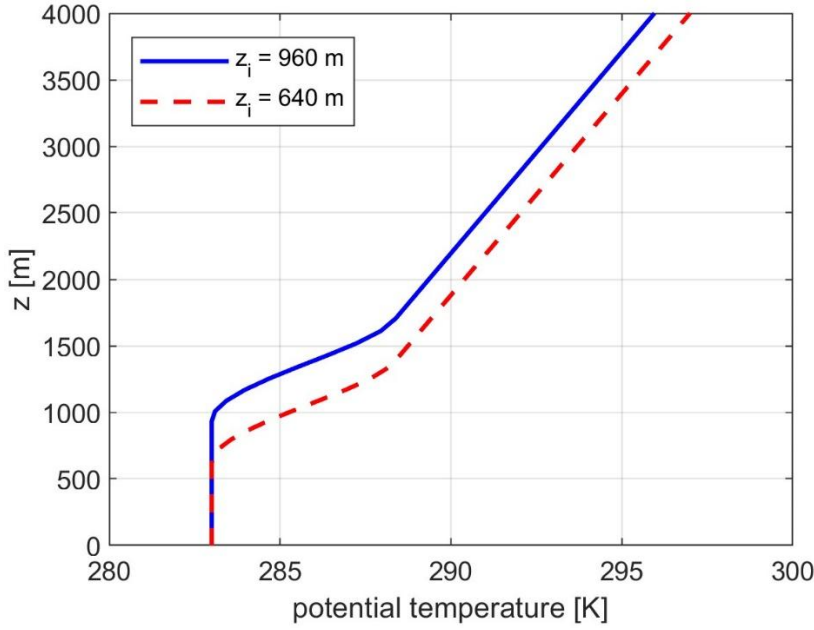


These curves, which are functions of hub-height freestream wind speed, must be converted to functions of  $U_{disk}$  to be used in the wind farm simulations. We make the conversion using a procedure similar to that described in van der Laan et al. (2014). It involves running single-turbine simulations in flat terrain over a set of wind speeds spanning the operating range of the wind turbine. In these simulations, where hub-height freestream wind speed is known, the actuator disk forces are set according to the original performance curves (Fig. 2). At the end of each simulation, we record  $U_{disk}$ . The outcome of this procedure is a set of curves for power, thrust, and rotor speed specified as functions of  $U_{disk}$ . These curves form a look-up table used to control the actuator disks during wind farm simulations and to support the post-processing of the results. We refer to this table as the turbine performance look-up table. This procedure is used to generate the turbine performance look-up tables for both the idealized and the real wind farm simulations.

The table includes an additional column, effective wind speed,  $U_e$ . For a given value of  $U_{disk}$  and corresponding values of thrust and power,  $U_e$  is the wind speed that yields the same power and thrust when using the original  $C_T$  and  $C_P$  curves. We use this quantity to help analyze the RANS results in Section 3.1.

## 2.4 Atmospheric conditions

The atmospheric conditions in the idealized wind farm simulations feature stable stratification overlying a neutral boundary layer. The stratification starts with a capping inversion approximately 600 m thick with a maximum potential temperature lapse rate of 10 K/km. The free atmosphere above has a potential temperature lapse rate of 3.3 K/km. Nearly all the conventionally neutral boundary layers simulated in this study use one of the two inflow potential temperature profiles depicted in Fig. 3. The primary difference between the profiles is  $z_i$ , the altitude at which the inversion starts. In one profile  $z_i$  is 640 m, in the other  $z_i$  is 960 m. We use a single-column (one-dimensional) RANS model to generate velocity and turbulence profiles that are in quasi-equilibrium with the prescribed potential temperature profile, given inputs such as latitude ( $= 43^\circ$ ), aerodynamic roughness length ( $= 0.028$  m), and geostrophic wind speed. In each case the geostrophic wind speed is set such that all the turbines in the simulated wind farm operate on the flat part of the  $C_T$  curve. The inflow profiles of velocity and turbulence are uniform above the inversion.



**Figure 3: Inflow vertical profiles of potential temperature.**

## 200 2.5 Mesh

The vertical spacing of the prismatic cells used to resolve the boundary and inversion layers relaxes from a cell height of 2 m at the ground to a cell height of 92 m or less at the top of the inversion layer. Above the inversion, the mesh further relaxes to isotropic cells with centroid spacings of approximately 200 m.

205 The horizontal mesh resolution is highest near the wind farm. The cell spacing is no more than 0.1 D within 12 rotor diameters of any turbine up to a height of 335 m. The cell spacing in close proximity to the actuator disks is 0.05 D.

## 2.6 Wind farm and turbine performance metrics

The ultimate aim of this study is to better understand whether and to what degree terrain influences wind farm and turbine performance. The primary performance metric for the wind farm is array efficiency, which is defined as follows:

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$$\eta = \frac{\sum_{k=1}^N P_{WF,k}}{\sum_{k=1}^N P_{I,k}} \quad (2)$$

$P_{WF,k}$  is the power from turbine  $k$  when simulated in the wind farm, and  $P_{I,k}$  is the power from turbine  $k$  when simulated in isolation. The terms *array efficiency* and *wind farm efficiency* are used interchangeably in this paper. We also refer to  $\eta$  as a *turbine interaction loss factor*. The percent turbine interaction loss,  $L$ , is simply  $100\% \times (1 - \eta)$ .



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Reliably calculating the values in Eq. (2) requires results from three types of simulations: wind farm (all actuator disks on), freestream (all actuator disks off), and isolated turbine (one actuator disk on). All three types are run for each combination of simulated terrain, atmosphere, and turbine layout – using the same domain, mesh, and boundary conditions. The wind farm simulation yields the  $P_{WF,k}$  values. To get the  $P_{I,k}$  values, we could run an isolated turbine simulation at each turbine location in the wind farm layout, but this is not necessary. Instead, we run a subset of the turbines in isolation, typically four turbines. Using results from the freestream simulation together with the isolated turbine results, we can determine the relationship between the freestream conditions and  $U_{disk}$  for the isolated turbines. This enables reliable estimation of  $U_{disk,I,k}$  and, in turn,  $P_{I,k}$  for every turbine. As will be discussed in Section 3.3, the relationship between freestream conditions and  $U_{disk,I,k}$  can be sensitive to the position of the turbine relative to the hill profile, so when simulating two rows on a hill, two sets of isolated turbine simulations are needed, one for each row.

The primary turbine performance metric is power coefficient:

$$C_P = \frac{P_I}{\frac{1}{2} \rho A U_{\infty,disk}^3} \quad (3)$$

$P_I$  is the power from a turbine simulated in isolation.  $\rho$  is the density, and  $A$  is the swept area of the rotor.  $U_{\infty,disk}$  is the axial component of velocity averaged over the disk in the freestream simulation.  $C_P$  values reported herein correspond to an average of the isolated turbine simulations run for any given terrain-atmosphere-turbine-row combination.

The power values used in Eq. 2 and Eq. 3 are not the dot product of thrust and velocity at the actuator disk; the power instead derives from  $U_{disk}$  and the turbine performance look-up table.

### 3 Results

Here we present the simulation results, dividing them into four sections. Results from the idealized two-row wind farms and idealized single-row wind farms are in sections 3.1 and 3.2, respectively. These sections focus on the impact of terrain on wakes, wind farm blockage, and turbine interaction loss. The third section, 3.3, presents results from turbines simulated in isolation and focuses on the impact of terrain on the near-wake, turbine blockage, and turbine efficiency ( $C_P$ ). Results from simulations of a real wind farm are in Section 3.4.

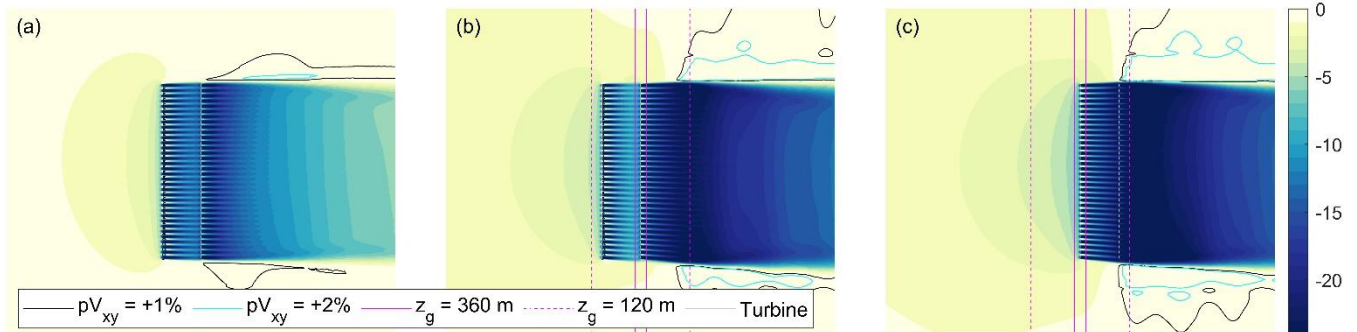


This results section includes many plots of and references to “wind speed” and “pressure”. Here, *wind speed* is the horizontal component of velocity. *Pressure* is relative pressure. This relative pressure is equal to the absolute pressure minus the pressure field in balance with both the geostrophic Coriolis force and the inflow potential temperature profile.

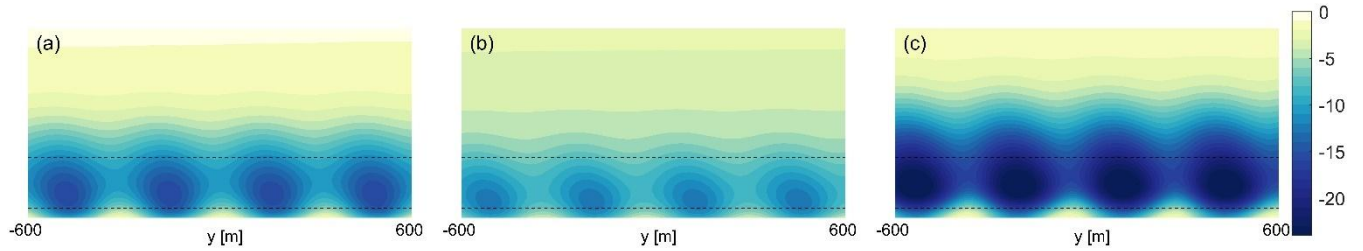
### 3.1 Two-row wind farms

We simulate three different two-row wind farms: *flat*, *hill windward*, and *hill leeward*. The turbines in each row are arranged south-to-north with the second row directly to the east of the first row. The x-component of the distance between the rows is 15 D. In the *flat* case, the wind farm operates in flat terrain. In the *hill windward* case, the first row is on the windward side of the hill and the second row is on the crest. In the *hill leeward* case, the first row is on the crest and the second row is on the leeward side of the hill. The *hill windward* and *hill leeward* cases are run with the same inflow conditions. In the freestream simulations, the hub-height wind speed is 9 m/s at the crest, 6.2 m/s 15 D upstream of the crest and 6.8 m/s 15 D downstream of the crest. In the *flat* case, the freestream hub-height wind speed is 9 m/s across all turbine locations. In all three cases,  $z_i = 960$  m.

Terrain has a significant impact on wake recovery and upstream blockage effects in these simulations. This can be seen in Fig. 4, which shows the percent change in wind speed at hub height due to the presence of the wind farm (flow is from left to right, as is the case in all top view plots in this paper). The wakes downstream of the *hill windward* and *hill leeward* wind farms are clearly deeper than in the *flat* case. In addition, the upstream slowdowns are more pronounced and far-reaching in the hill cases, especially the *hill leeward* case. Finally, a close look at Fig. 4 reveals that the wakes in between the two rows are deeper in the *hill leeward* case than in the *hill windward* case—with the flat case in between. The case-to-case differences in the wakes from the first row can be seen more clearly in Fig. 5, which shows the percent change in wind speed relative to freestream on a vertical plane 2.5 D upstream of the second row.



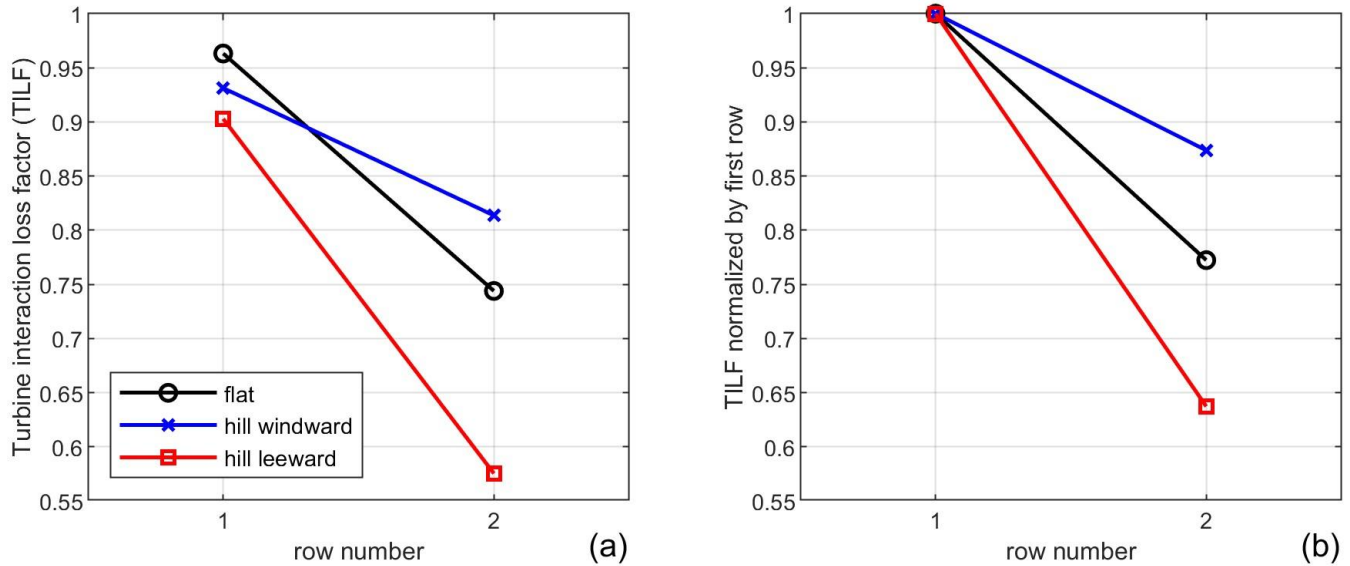
**Figure 4: Top view of the percent change in wind speed relative to freestream at hub height for the *flat* (a), *hill windward* (b), and *hill leeward* cases (c). Magenta lines are ground elevation contours; the hill crest is located between the two solid magenta lines.**



**Figure 5:** Percent change in wind speed relative to freestream on a vertical plane located 2.5 D upstream of the second row in the *flat* (a), *hill windward* (b), and *hill leeward* cases (c). Dashed lines demark the rotor layer (heights of 25 m and 195 m).

270 The wake and blockage trends in Fig. 4 are reflected in the row-by-row performance. This can be seen in Fig. 6. The first row in the *hill leeward* case is less efficient than the first row in the *hill windward* case, which is less efficient than that first row in the *flat* case. This order is consistent with the apparent strength of the upstream blockage effects in Fig. 4. The *hill leeward* case also has the least efficient second row; the *hill windward* case has the most efficient second row. The efficiency of the second row relative to the first row is clearly very sensitive to terrain as highlighted in Fig. 6b. The primary reason for

275 this efficiency trend is differences in streamwise pressure variation.



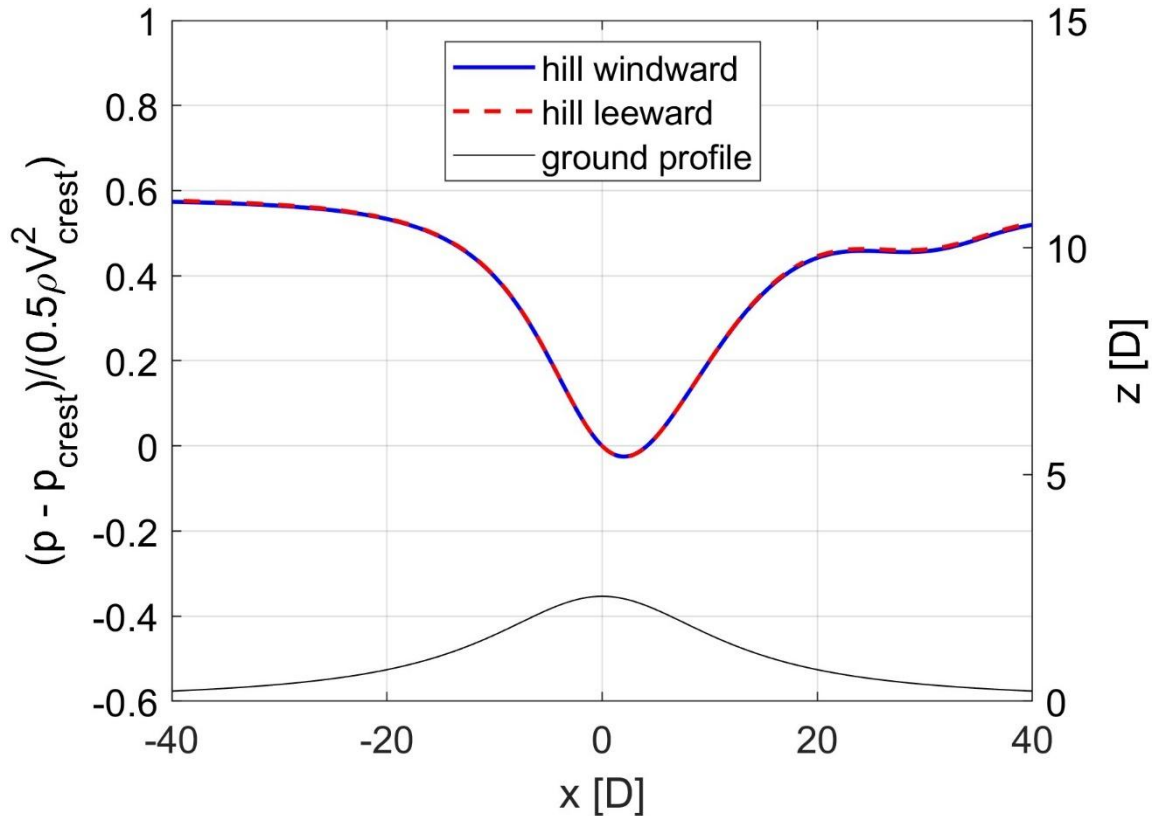
**Figure 6:** Turbine interaction loss factor,  $\eta$ , by row (a); turbine interaction loss factor normalized by the turbine interaction loss factor for the first row (b).

280 The hill induces substantial streamwise pressure gradients in the vicinity of the wind farm. Figure 7 shows how the freestream pressure varies along the hill profile (flow is from left to right, as is the case in all similar plots in this paper). The pressure gradient is favorable on the windward side of the hill and generally unfavorable (adverse) on the leeward side. As discussed in the Introduction, FPGs promote wake recovery; APGs do the opposite. The wakes from the first row in the *hill windward* case travel through a favorable pressure gradient before reaching the second row. The pressure drop between these rows accelerates wake recovery, and as a result, the second row experiences less turbine interaction loss. The wakes from the



285 first row in the *hill leeward* case travel primarily through an APG, deepening the wakes and increasing the turbine interaction loss in the second row. The freestream streamwise pressure gradient in the *flat* case is negligible, and the turbine interaction loss in the second row relative to the first naturally falls between the two other cases (Fig. 6b).

Terrain can also affect the trajectory of a wake. In the *hill leeward* case, the wakes from the first row are clearly higher than  
 290 in the other two cases (see Fig. 5). The wakes from the first row in the *hill windward* case are slightly lower than those in the *flat case*. These differences must have some effect on the turbine interaction loss trends in Fig. 6, but the effect is probably small compared with the effect of the wake deficit differences evident in Fig. 5. The effect of terrain on wake trajectory may be more important at wind farms located in more complex terrain.



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**Figure 7: Freestream pressure coefficient along the hill profile at hub height in the middle of the domain ( $y = 0$  m) for the *hill windward* (blue) and *hill leeward* (red) cases. The pressure coefficient is relative to conditions at the hill crest. The thin black line is the hill profile.**

We also calculate turbine interaction loss,  $100\% \times (1 - \eta)$ , for the full wind farm. The turbine interaction loss is 14.6% in  
 300 the *flat* case. It is 15.8% in the *hill windward* case and 18.5% in the *hill leeward* case. In a typical energy yield analysis, the turbine interaction loss for a wind farm in terrain would be estimated using an engineering model that does not include the



effect of terrain on wakes and blockage. If we were to neglect the influence of ground elevation variation in this RANS-based analysis and use the *flat* case result ( $L = 14.6\%$ ) to predict the losses for the two hill cases, the losses would be underestimated, especially for the *hill leeward* case. That said, in a typical energy yield analysis, the turbine interaction loss calculation would consider how the freestream wind speed varies over the terrain.

There does not appear to be a standard approach for incorporating terrain-induced spatial wind speed variations in a wind farm flow analysis run with a model that does not include terrain. Here we explore one plausible hybrid approach, using the freestream RANS solution for the quasi-2D hill in combination with the RANS solution for the two-row wind farm in flat terrain. From the *flat* case analysis, we can determine the effective wind speed experienced by each turbine relative to the effective wind speed each turbine would experience operating in isolation  $(U_e/U_{e,I})_{flat}$ . We determine  $U_{e,I,hill}$  for each turbine from the freestream and isolated turbine simulations run with the hill terrain as described in Section 2.6. Multiplying the two sets of values yields  $U_{e,hill estimate}$  for each turbine. As explained in Section 2.3, the  $U_e$  values can be used to look up power. These power values yield hybrid hill-flat turbine interaction loss predictions of 20.3% for the *hill windward* case and 9.6% for the *hill leeward* case. These hybrid predictions are much different from the loss values derived from simulations of the wind farms with the terrain included. Not only are the errors large, but the windward-to-leeward loss trend is dramatically reversed. This hybrid approach does not appear to be a good option for layout optimization here.

At first glance, the huge differences between the hybrid hill-flat predictions and the full wind-farm-terrain results seem hard to believe; however, the errors from the hybrid approach make more sense after considering how the wind speed varies across the hill profile in conjunction with the row-by-row efficiency numbers in Fig. 6a. The freestream wind speeds in the *hill windward* case are much higher at the location of the second row, on the crest, compared with the first row—with the same being true for  $U_{e,I}$  and  $P_I$ . As a result, the efficiency of the second row has a much greater influence on the overall efficiency than the efficiency of the first row. Thus in this hybrid approach, where the efficiencies come from the *flat* case, the low efficiency of the second row (74.4%) dominates the calculation of the overall wind farm efficiency, which in this hybrid hill-flat approach is calculated to be 79.7% – a turbine interaction loss of 20.3%. Of course in the actual *hill windward* case, the second-row efficiency also makes the largest contribution to the overall efficiency, but in this case the second-row efficiency is quite a bit higher (81.4%), resulting in a comparatively lower overall loss of 15.8%. In the *hill leeward* case, the  $U_{e,I}$  and  $P_I$  values in the first row, on the crest, are much higher than those in the second row downstream. As a result, the first row makes a much larger contribution to the overall wind farm efficiency/loss. The hybrid flat-hill prediction of the overall loss for the *hill leeward* case effectively weighs the 3.7% loss in the first row much more heavily than the 25.6% loss in the second row (both numbers from the *flat* case), yielding an overall loss prediction of just 9.6% which is quite a bit lower than the 18.5% overall loss calculated directly from the actual *hill leeward* analysis.



335 We have not explored other methods used to combine freestream flow over terrain with wind farm modeling without terrain.  
 Like with wake superposition, there are likely several different methods currently in use, but unlike with wake superposition,  
 these methods are not to our knowledge well documented in the literature. The hybrid approach used here seems reasonable,  
 and intuitively one might expect it to at least offer an improvement over methods that completely ignore terrain-induced  
 wind speed variations when predicting turbine interaction losses; however, in these examples, the hybrid approach makes the  
 340 loss predictions much worse. For a hybrid approach to consistently work well, it would probably need to account for the fact  
 that streamwise pressure gradients affect low-momentum wake flow more than higher-momentum freestream flow.

### 3.2 Single-row wind farms

We ran 28 single-row RANS-based wind farm flow analyses, varying row location, terrain, and inflow conditions. Table 1  
 lists the high-level results for twelve of these cases. The turbine interaction loss,  $L$ , varies a lot across these single-row wind  
 345 farms, from -1.8% to +15.6%. The following pages and figures explore the RANS results in more detail in order to better  
 understand this variation and how it relates to wakes, blockage, and terrain.

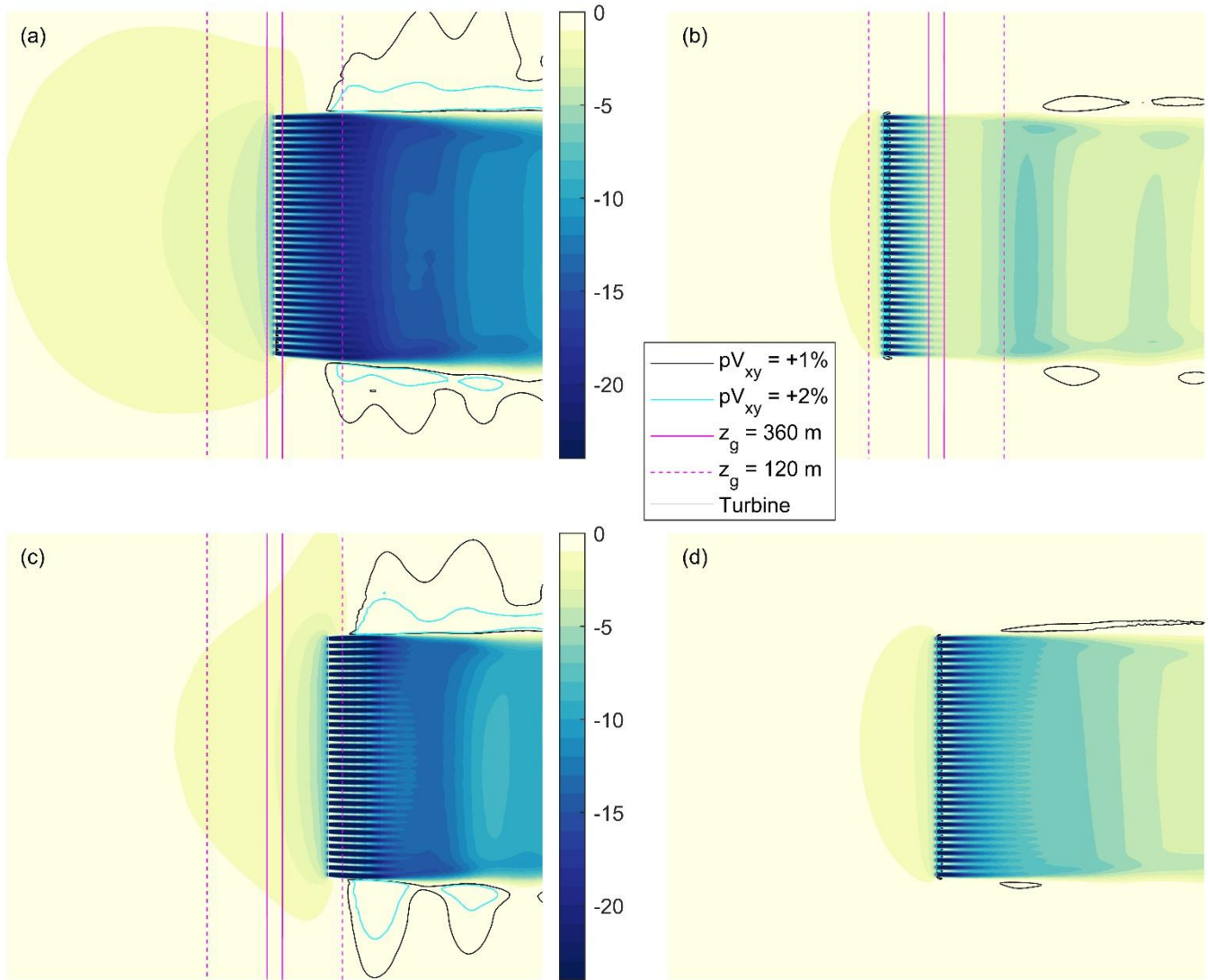
Case	Terrain	Row location	$z_i$ [m]	$U_{\infty, disk}$ [m/s]	$L$ [%]	$C_p$
A	Hill	crest	960	9.0	6.94	0.446
B	Hill	windward	960	8.1	-0.27	0.456
C	Hill	leeward	960	9.0	9.31	0.421
D	Flat	N/A	960	8.9	1.46	0.444
E	Hill	crest	640	9.0	10.75	0.445
F	Hill	windward	640	7.7	-0.01	0.456
G	Hill	leeward	640	8.6	15.61	0.408
H	Flat	N/A	640	9.0	3.23	0.442
I	Narrow Hill	crest	960	9.4	8.97	0.437
J	Valley	leeward	640	6.3	-1.80	0.477
K	Hill	leeward	960	6.5	3.57	0.421
L	Valley	leeward	640	8.0	1.38	0.461

**Table 1: Case descriptions and performance values from twelve different single-row wind farm analyses**

We first look at cases A-D, which are all simulated with a  $z_i$  of 960 m. Differences in wake and blockage effects across these  
 cases can be seen in Fig. 8, which shows top views of the percent change in wind speed relative to freestream conditions at  
 350 hub height. The wake recovery and upstream blockage effects are much different in the cases with the hill compared with the  
 flat terrain case. When the turbine row is on the crest or on the leeward side of the hill (Fig. 8a and Fig. 8c), the upstream

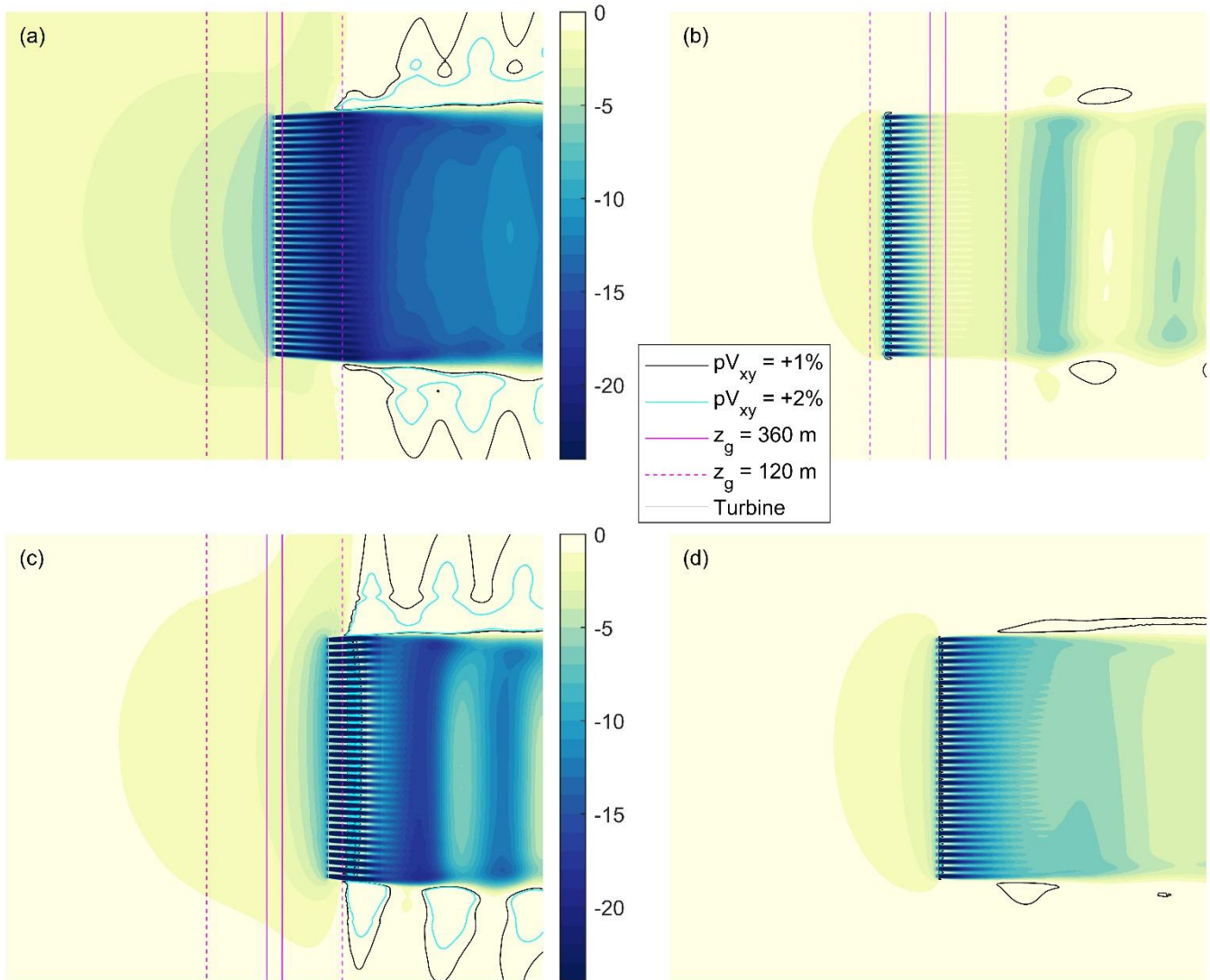


blockage effects are much more pronounced than in the flat case (Fig. 8d), and the wind farm wake downstream is deeper. In contrast, when the turbine row is on the windward side of the hill (Fig. 8b), the wind farm wake deficit is smaller than in the flat case and the upstream blockage effect is less pronounced.



**Figure 8: Top view of the percent change in wind speed relative to freestream at hub height for Case A (a), Case B (b), Case C (c), and Case D (d) as defined in Table 1. Magenta lines are ground elevation contours; the hill crest is located between the two solid magenta lines.**

The Case A results relative to Case D are consistent with the RANS results in the master's thesis of Kashyap (2022), which found that the upstream blockage effect of an infinitely long row of wind turbines can be stronger when the row is located along the crest of a quasi-2D hill as compared to when located in flat terrain.

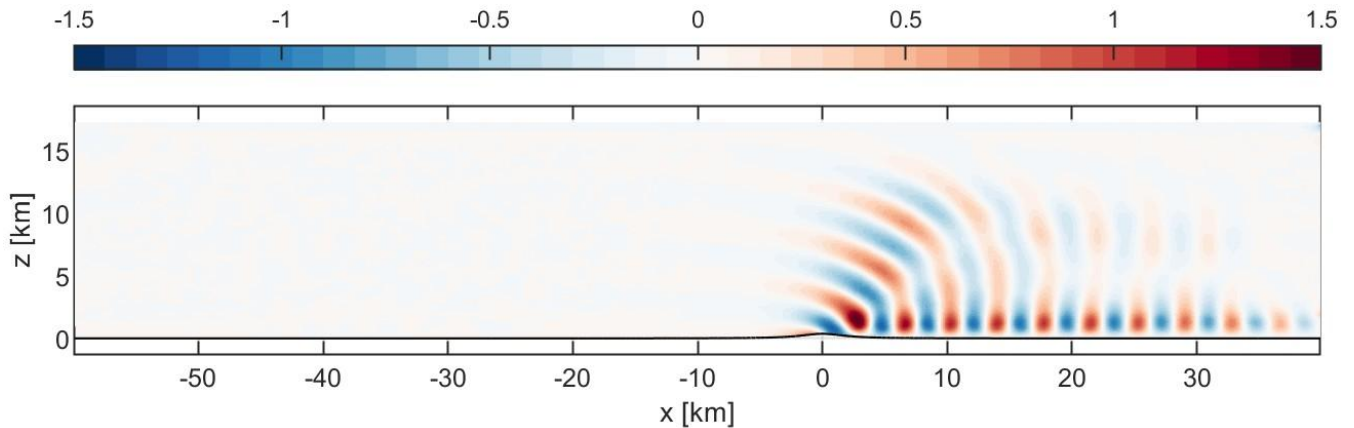


365 **Figure 9: Top view of the percent change in wind speed relative to freestream at hub height for case E (a), case F (b), case G (c), and case H (d) as defined in Table 1. Magenta lines are ground elevation contours; the hill crest is located between the two solid magenta lines.**

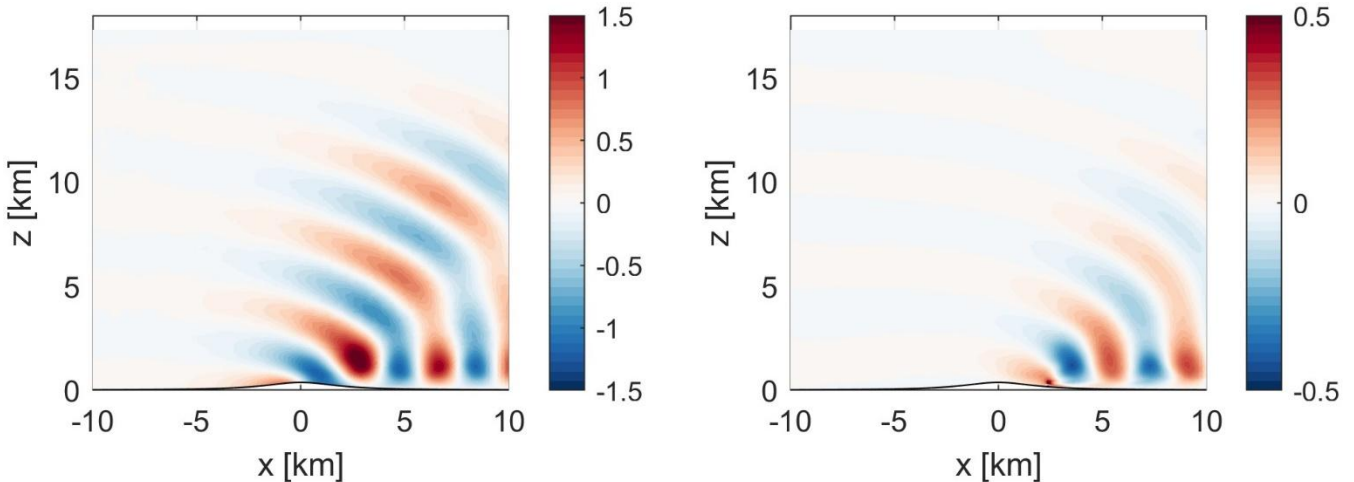
Figure 9 shows the percent change in wind speed relative to freestream for cases E-H; these cases are similar to cases A-D but with  $z_i = 640$  m. Compared with the  $z_i = 960$  m results in Fig. 8, the case-to-case trends in wake and blockage effects in  
 370 Fig. 9 are similar but more pronounced—as are the case-to-case loss trends (compare cases E-H with cases A-D in Table 1). The effect of gravity waves is also more clearly evident in Fig. 9, especially for the case with the row on the leeward side of the hill (Case G). Fig. 10 shows the vertical component of velocity in a side view of the freestream solution for this case.



Hill-induced gravity waves can be seen in the figure. Despite the large amplitude of these waves, there is no indication of gravity waves reflecting off a boundary and polluting the results in this case or in any of the other cases in this study.



**Figure 10: Case G, side view of the vertical component of freestream velocity [m/s] near the middle of the domain ( $y = 160$  m).**

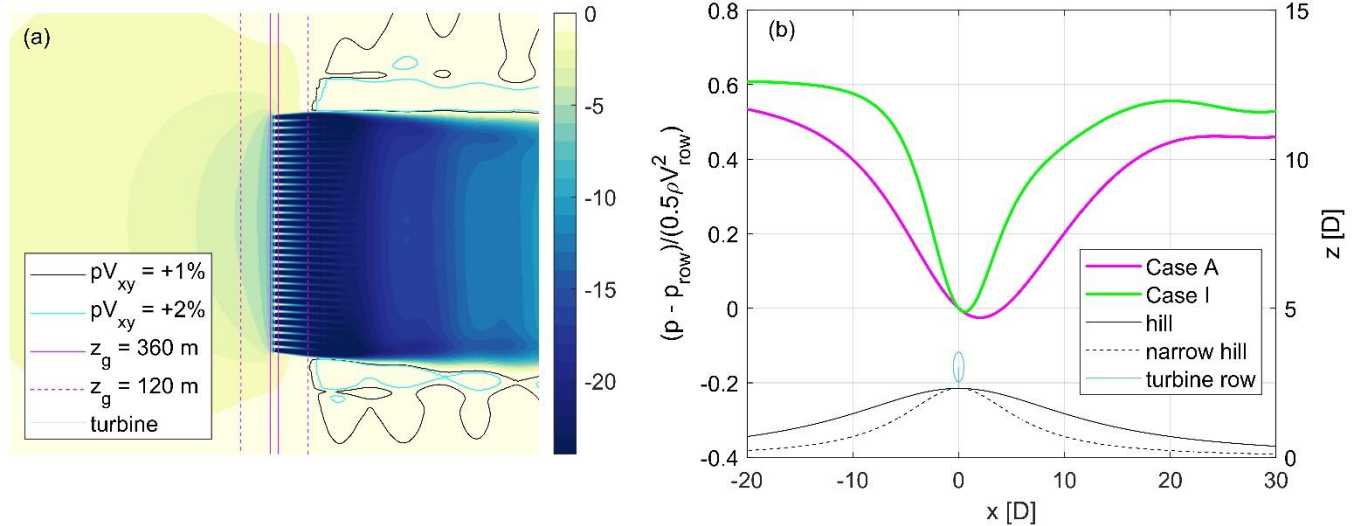


**Figure 11: Case G, side view of the vertical component of velocity [m/s] near the middle of the domain ( $y = 160$  m): freestream solution (a); wind farm solution minus freestream (b).**

The wind farm also induces gravity waves, a finding consistent with Allaerts and Meyers (2018). Figure 11a is a reproduction of Fig. 10 but zoomed in; Fig. 11b is from the same case but shows the wind farm solution minus the freestream solution, revealing the effect of the wind farm on the vertical component of velocity. The gravity waves induced by the wind farm have similar characteristics to the gravity waves induced by the hill, but with a lower amplitude and, because the wind farm is  $15 D$  downstream of the crest, a different initiation point and phase.



Differences in the upstream blockage effects evident in Fig. 8 and Fig. 9 are consistent with the differences in turbine interaction loss across the same eight cases in Table 1. The upstream blockage effects are, of course, the cause of turbine interaction loss in single-row wind farms. The rest of this subsection focuses on *how* terrain modulates these upstream effects.



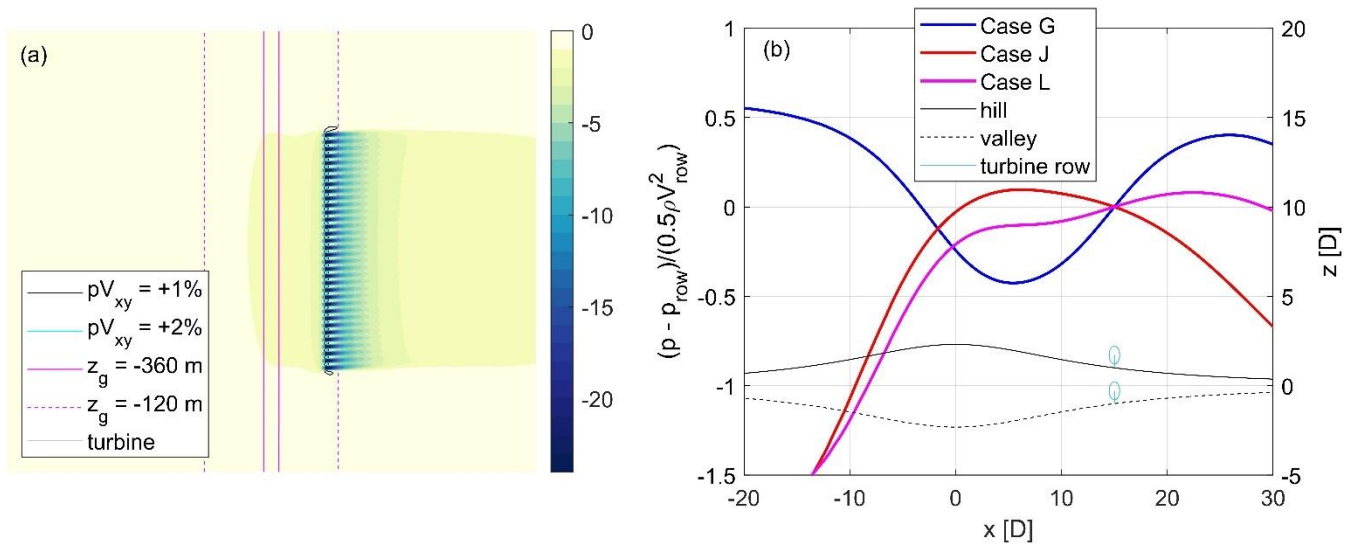
**Figure 12: Top view of the percent change in wind speed relative to freestream at hub height for Case I (a), and freestream pressure coefficient along the hill profiles at hub height at  $y = 0$  m (b). Case I (narrow hill) is in green. Case A (hill) is magenta. The pressure coefficient is relative to conditions at the hill crest, where the wind turbines in the corresponding wind farm simulations are located. The thin black lines depict the two hill profiles.**

We begin this effort with an analysis featuring a hill that is narrower than the baseline hill. Case I is similar to Case A ( $z_i = 960$  m with the turbines on the hill crest), but it has a hill profile that is half as wide. Figure 12a shows the percent change in wind speed relative to freestream for Case I, and Fig. 12b shows the streamwise freestream pressure variation for both the narrow hill case (Case I) and the similar baseline hill case (Case A). Compared with the baseline hill, there is a sharper pressure rise downstream of the crest of the narrow hill case. This deepens and expands the wake downstream, increasing the wake blockage, which appears to contribute to an increase in the upstream blockage effect and, in turn, a larger loss for the narrow hill case.

The concept of wake blockage is well known among those who conduct wind tunnel experiments, but it is seldom discussed in the wind energy community. Blockage is an obstruction causing oncoming flow to diverge from the path it would otherwise take. The obstruction need not be solid – a local region of relatively slow-moving flow can have a similar effect. In this discussion, the slow-moving flow feature is the wind farm wake. According to the continuity equation, the presence of the wind farm wake means that flow must diverge and slow upstream. Because hill-induced streamwise pressure gradients affect the wind speed in the wake and the size of the wake, it also affects how much oncoming flow is deflected relative to freestream conditions. This deflection cannot happen discontinuously. It happens over a distance, a distance that could



include locations upstream of the wind farm. In other words, according to this theory, wake blockage, which can be modified by streamwise pressure gradients, influences the blockage-related slowdown upstream of the wind farm. These slowdowns in turn affect the wind speeds experienced by the turbines and the turbine interaction loss. A similar theory was put forward in Revaz and Porté-Agel (2024) to explain the effect of streamwise pressure gradients on the induction of an individual turbine; however, to our knowledge, the theory has not, until now, been applied to wind farm blockage in terrain. This theory is not a complete explanation: there are likely other terrain-induced influences on upstream blockage effects. Nevertheless, it does appear to explain some of the terrain-related blockage trends in the single-row results.

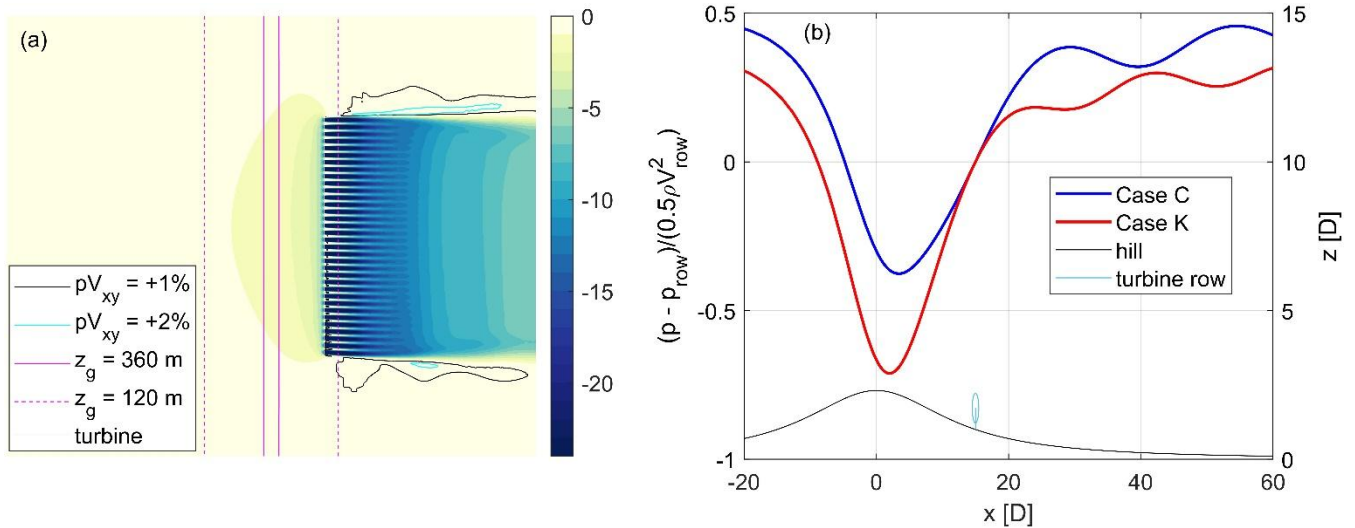


**Figure 13: Top view of the percent change in wind speed relative to freestream at hub height for Case J (a), and freestream pressure coefficient along the terrain profiles at hub height at  $y = 0$  m (b). The Case J (valley) pressure coefficient is in red; Case G (hill) is in blue; and Case L (valley) is magenta. The pressure coefficient is relative to conditions 15 D downstream of the hill crest and valley trough, the location of the turbine rows in the corresponding wind farm simulations. The thin black lines depict the valley and hill profiles.**

We also analyzed two valley cases, Case J and Case L. The valley is a mirror image of the baseline hill and the row is on the leeward side of the valley, 15 D downstream of the minimum elevation of -370 m. The altitude where the inversion starts,  $z_i$ , is 640 m. The main difference between the two cases is the prescribed geostrophic wind speed, which is 13.4 m/s for Case J and 16.2 m/s for Case L. Figure 13a shows the percent change in wind speed relative to freestream for Case J, and Fig. 13b shows the streamwise variation of freestream pressure at hub height along the terrain profiles for the two valley cases as well as for the hill case with the same  $z_i$  and turbine row location (Case G). The background streamwise pressure gradient near the turbine locations in Case J is clearly favorable, in sharp contrast to the hill case. This results in a huge difference in the wake recovery (compare Fig. 13a with Fig. 9c). Again, we believe this difference in the wakes feeds back and contributes to significant differences in the effect of blockage upstream. The difference in blockage-related upstream slowdowns causes large differences in the turbine interaction loss: 15.6% for the hill case and -1.8% for this valley case, a turbine interaction gain. The faster flow in Case L responds differently to the terrain compared with Case J, resulting in a mild adverse pressure



435 gradient in the background flow near the turbines (Fig. 13b) and a higher turbine interaction loss (+1.4% vs. -1.8%). The difference in freestream streamwise pressure gradients across the three cases also contributes to differences in the turbine efficiency ( $C_p$ ) as discussed in Section 3.3.



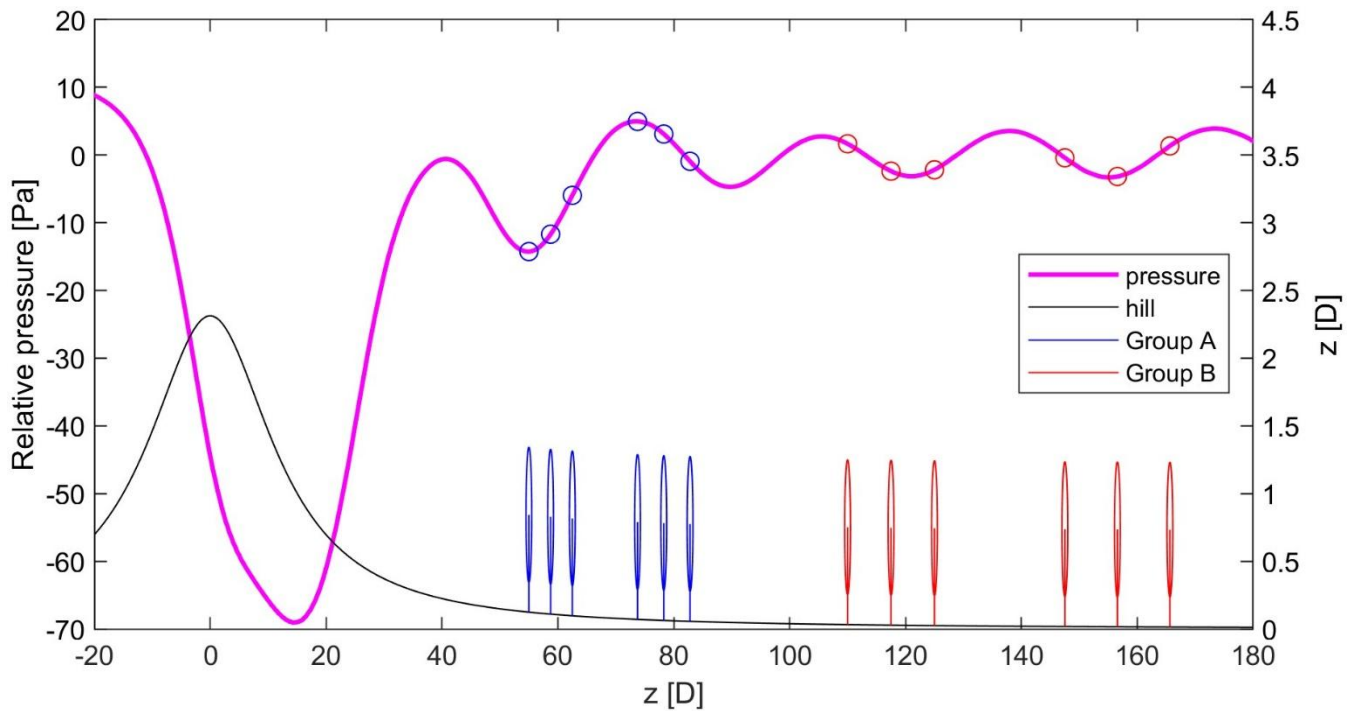
440 **Figure 14: Top view of the percent change in wind speed relative to freestream at hub height for Case K (a), and freestream pressure coefficient along the hill profiles at hub height at  $y = 0$  m (b). Case K is in red, Case C is in blue. The pressure coefficient is relative to conditions 15 D downstream of the hill crest, the location of the turbine rows in the corresponding wind farm simulations. The thin black line corresponds to the hill profile.**

The effect of inflow wind speed on turbine interaction loss is also evident in the hill simulations. Case K is similar to Case C  
 445 but runs with a lower geostrophic wind speed of 6.6 m/s compared with 8.0 m/s in Case C. This difference in wind speed results in a very different predicted loss (see Table 1). Figure 14a shows percent change in wind speed relative to freestream for Case K, and Fig. 14b shows the streamwise variation of freestream pressure along the hill for Case K and Case C. The phase and wavelength of the freestream gravity waves in the lee of the hill are clearly different between these two cases, with the lower wind speed case exhibiting hill-induced gravity waves with longer wavelengths. The differences in the gravity  
 450 waves contribute to differences in the streamwise variation of pressure in the lee of the hill. The streamwise pressure gradient at the turbine row is about the same between the two cases, but there is a greater pressure rise further downstream in Case C. This will contribute to a deeper wake. The fact that the wind farm wake in Case K is not as deep or thick compared to Case C may contribute to differences in the upstream blockage effect. In this way, hill-induced gravity waves indirectly contribute to upstream slowdowns. Gravity waves, and more generally the response of stratified flow to the hill and the  
 455 turbines, may also affect these slowdowns more directly – as will be discussed later in this section.

We ran additional simulations not listed in Table 1 in order to further explore the various influences on turbine interaction loss in single-row wind farms. Twelve of the additional single-row analyses were run using the Case F inflow conditions and



terrain. Gravity waves in this case induce significant streamwise pressure variations, even on the relatively flat terrain downstream of the hill. This can be seen in Fig. 15, which also shows the locations of the 12 additional single-row wind farms. The Group A wind farms correspond to locations where there are large local variations in the background pressure. The Group B wind farms are further downstream and generally correspond to locations where the local variations in pressure are smaller.



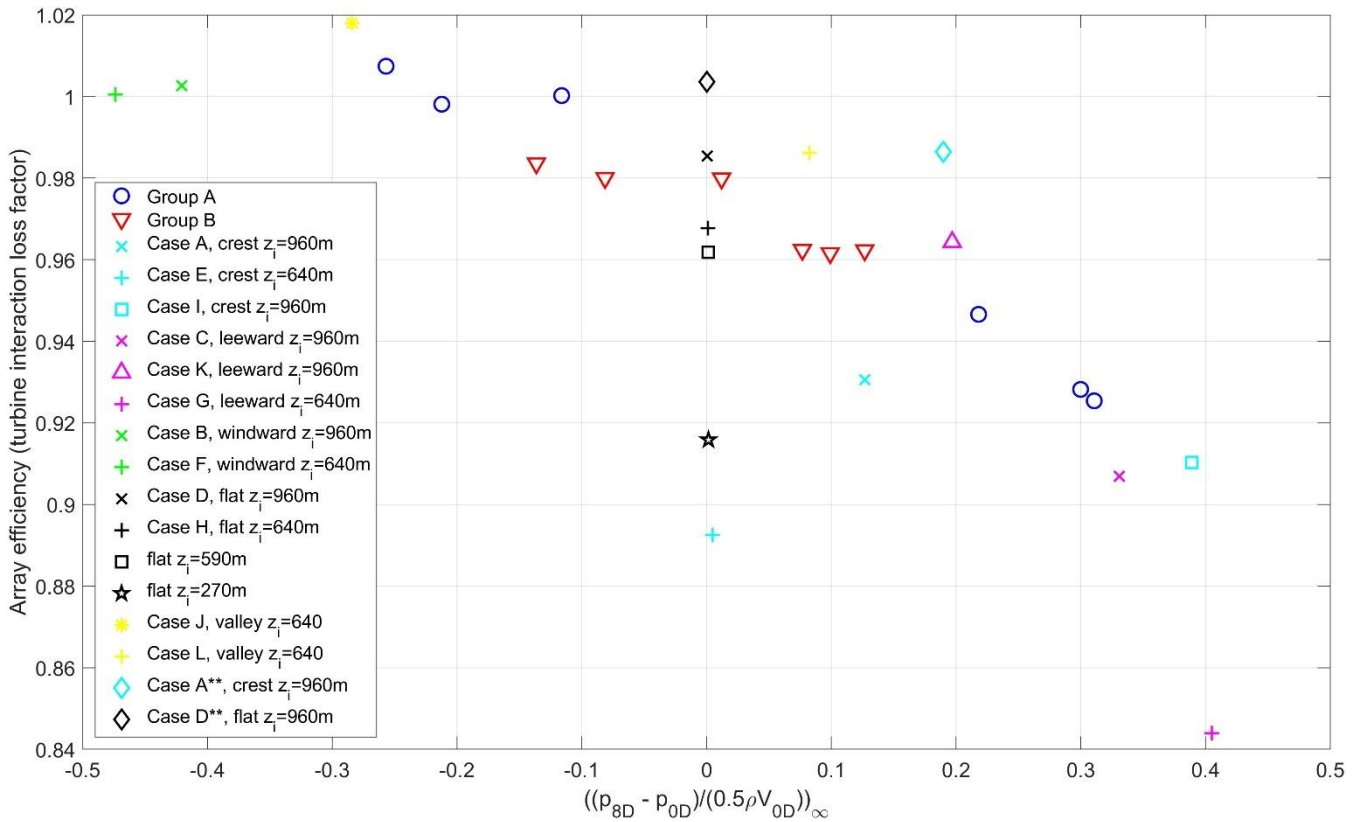
**Figure 15: Freestream pressure variation for Case F. The locations of the six single-row wind farms in Group A (blue) and the six single-row wind farms in Group B (red) are shown relative to the location of the hill profile (black).**

Also, two additional analyses were run on flat terrain, each with a lower inversion height ( $z_i = 590$  m and  $z_i = 270$  m). Finally, Case A and Case D were rerun with the buoyancy term in the vertical momentum equation zeroed out.

Plotting array efficiency vs freestream pressure coefficient for all these cases provides insight into the physical influences on blockage-related upstream slowdowns. Pressure coefficient in Fig. 16 is equal to the pressure difference between the turbine row location and  $8D$  downstream, normalized by the dynamic pressure at the row location—with all values coming from the freestream simulation. There is a correlation between freestream pressure coefficient and array efficiency in Fig. 16. The correlation coefficient is  $-0.74$ , which translates to an  $r^2$  of  $0.54$  for a linear regression of the data ( $r^2$  is  $0.61$  when the two results without buoyancy are removed). The correlation does not materially improve when using different locations to calculate the pressure coefficient.



The trend across the flat-terrain cases (black symbols) is different from the overall trend in that there is significant variation in loss across the flat cases despite hardly any variation in the freestream pressure coefficient. As the inversion gets lower in the flat cases, the slowdown upstream of the single row becomes more pronounced and the array efficiency decreases, a trend consistent with the results reported in Bleeg and Montavon (2022). This finding suggests that there may be more influencing upstream slowdowns in single-row wind farms than just streamwise pressure variation in the background flow and its impact on wakes.



**Figure 16: Array efficiency vs. freestream pressure coefficient for all single-row cases. The pressures are extracted from the freestream simulation at the turbine row location and 8 D downstream at hub height at  $y = 0$  m. Cases without buoyancy are marked with two asterisks in the legend.**

The trend across the flat cases may partially explain the trend across the crest cases (light blue), which also deviates from the overall trend. For example, in Case E, where the row is on the crest and  $z_i$  is 640 m, the freestream pressure drop between the row location and 8 D downstream is negligible, yet the upstream blockage effect is very strong (Fig. 9a), and the array efficiency is lower than the efficiencies in all but one of the other cases. Note that in this case the height of the inversion above the ground at the hill crest is comparable to the height of the inversion in the flat case with  $z_i = 270$  m. The physical



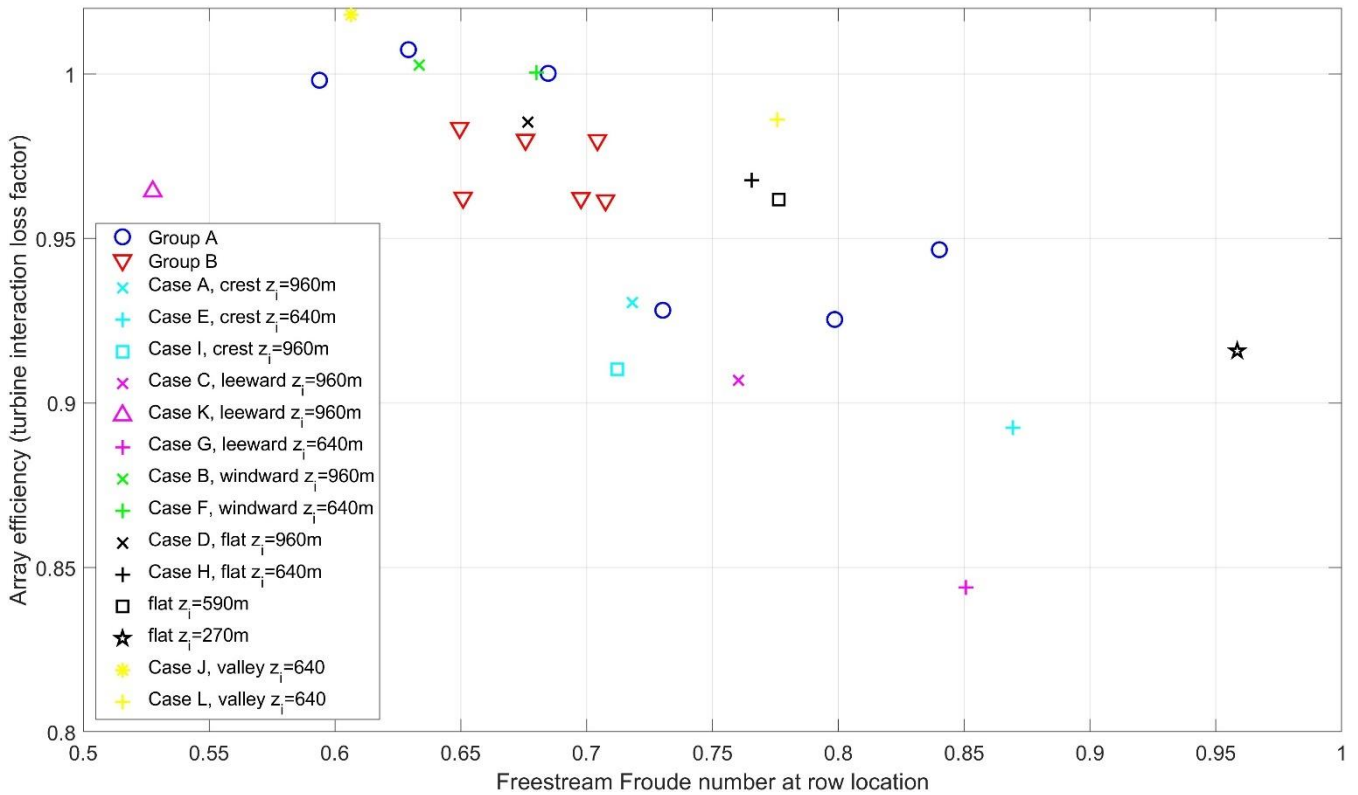
495 processes that contribute to the large upstream blockage effects in the flat case with  $z_i = 270$  m may also contribute to the large blockage effects in Case E. These processes probably influence the blockage effects in the other cases too.

These physical processes likely relate to the response of stratified flow to an obstacle, which in this study is a row of turbines. A non-dimensional number relevant to upstream blockage effects is the local Froude number, which we define as  
500 follows:

$$F = \frac{\bar{u}_{BL}}{\sqrt{g' h_{inv}}} \quad (4)$$

$h_{inv}$  is the vertical distance between the ground and the altitude where the vertical gradient of the potential temperature is at a maximum.  $\bar{u}_{BL}$  is the average of the x-component of velocity below  $h_{inv}$ . Finally,  $g'$  is the reduced gravity defined as  $g' =$   
505  $g \Delta\theta/\theta_0$ , where  $g$  is the acceleration due to gravity ( $9.81 \text{ m/s}^2$ ),  $\Delta\theta$  is the change in potential temperature across the inversion (5 K), and  $\theta_0$  is the reference potential temperature (286 K). The inputs to Equation 4 come from the freestream simulations.

The Froude number as defined above represents the ratio of the local wind speed to the wave speed along the inversion. As  
510 the freestream local Froude number increases and approaches 1.0 in these simulations, the upstream effect of a turbine row introduced to the flow generally increases, causing the efficiency of the row to decrease. This can be seen in Fig. 17, which shows array efficiency vs. local Froude number for all the cases except for the two cases with buoyancy zeroed out of the vertical momentum equation. There is a strong correlation between Froude number and single-row efficiency for the flat cases (black symbols). Across the other cases a correlation is still present, but it is not as strong. The correlation coefficient,  
515  $r$ , across all cases is -0.68. A multi-variable linear regression of the wind farm efficiency data with local Froude number and pressure coefficient as independent variables yields an  $r^2$  value of 0.76, which is significantly higher than the  $r^2$  values from separate single-variable linear regressions of the efficiency data ( $r^2$  is 0.47 when using local Froude number and 0.61 when using pressure coefficient). The efficiency trends may correlate better with different or additional independent variables. For example, vertical shear likely influences  $\eta$  as it did in Bleeg and Montavon (2022). In any case, the response of the stratified  
520 flow to terrain and the turbine row appears to have a significant influence on the upstream blockage effect in these simulations. Underlining this point, if we turn off the buoyancy in cases A and D (diamond symbols in Fig. 16), the upstream blockage effect decreases and the row efficiency increases, especially for Case A, where the row is on the hill crest. That said, even with the buoyancy turned off, blockage effects and array efficiency are still sensitive to terrain (the  $\eta$  difference between the two cases is 1.92%). So while buoyancy is a primary contributor to the effect of terrain on blockage in this  
525 study, it clearly is not the only contributor.



**Figure 17: Array efficiency vs. local freestream Froude number for all single-row cases with buoyancy in the momentum equation.**

### 3.3 Turbine performance

Not only can terrain influence wind farm efficiency, it can also affect turbine efficiency. The nominal  $C_P$  level for the simulated turbine is 0.445 when the hub-height freestream wind speed is between 6.1 m/s and 9.6 m/s (Fig. 2). Even though the simulated turbines in this study operate within this wind speed range, the  $C_P$  values derived from the results using Eq. (3) can depart significantly from 0.445. In Table 1, the  $C_P$  values vary between 0.408 and 0.477.

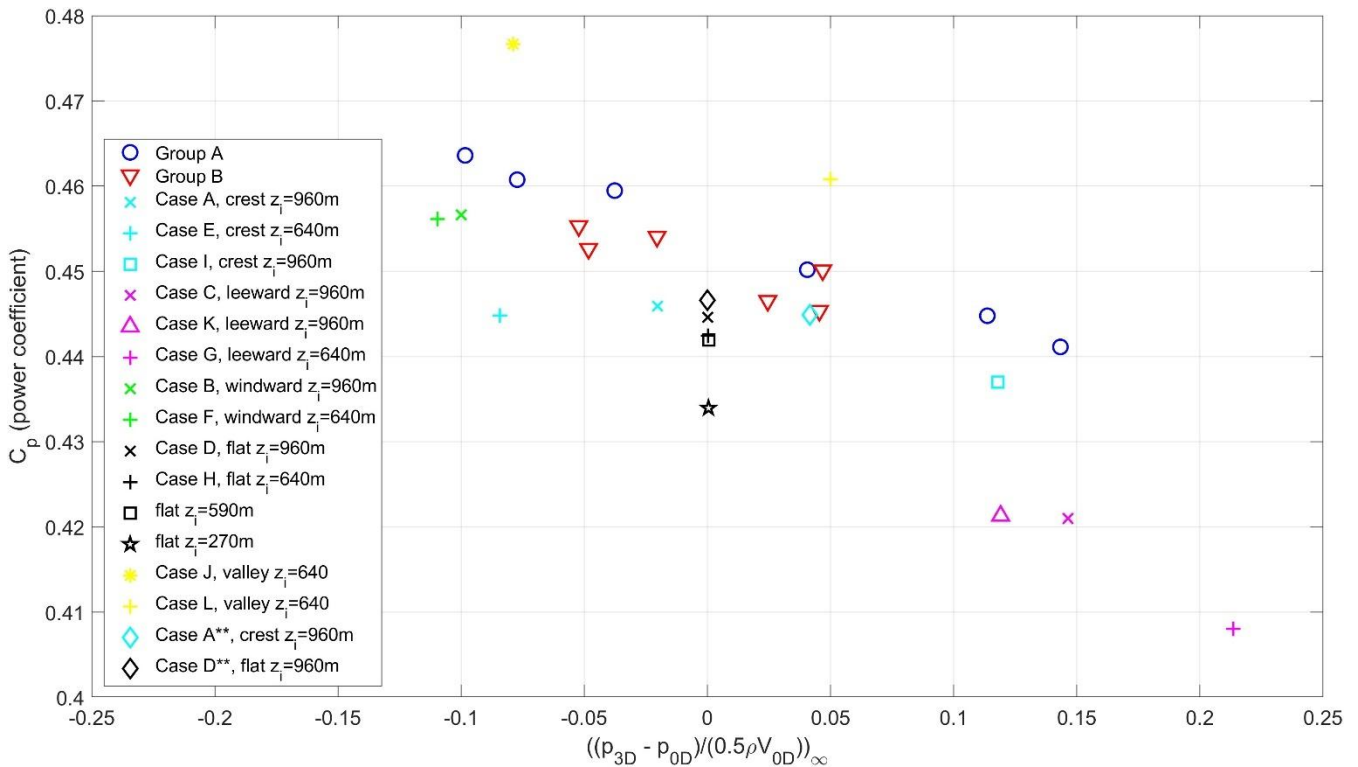
Turbine power is traditionally seen as a function of freestream conditions, but in fact it is more directly a function of the wind conditions at the rotor face during turbine operation. It is the conditions at the rotor face that determine the loads on the blades and in turn the rotor thrust and torque. The actuator disks in this study behave similarly: thrust and torque are functions of wind conditions at the disk when the actuator disk is on. The power coefficient,  $C_P$ , is the ratio between the power of a solitary turbine and the flux of kinetic energy passing through the rotor area when the turbine is not there.  $C_P$  varies substantially across the different cases in this study, implying that terrain can change the relationship between the freestream conditions and the conditions at the rotor disk. More specifically, it can affect induction factor. If terrain leads to an increase in the induction factor, the wind speed at the turbine rotor face decreases, decreasing power relative to the freestream kinetic energy flux. In other words, for a given turbine model an increase in induction factor reduces  $C_P$ .



545 Variations in induction factor and power coefficient are perhaps more commonly associated with wind turbine design changes. Here, rather, we claim that for a given turbine design, external factors, such as terrain-induced streamwise pressure variations, can affect induction factor and thereby  $C_p$ . Figure 18 plots  $C_p$  against freestream pressure coefficient for the cases in this study. This time the pressure coefficient is between the turbine row locations and locations 3 D downstream (instead of 8 D). There is a correlation between  $C_p$  and the freestream pressure coefficient ( $r = -0.77$ ), suggesting that streamwise

550 variation in the background freestream pressure across the near wake may influence  $C_p$ . We believe the mechanism is similar to that described in the previous section and in Revaz and Porté-Agel (2024): streamwise variations in the background pressure affects the depth and size of the near wake and in turn the amount by which oncoming flow deflects relative to freestream conditions. The amount by which the oncoming flow deflects directly connects with induction and thereby  $C_p$ . As discussed in the Introduction, many other researchers have identified a connection between streamwise variation in the

555 background (freestream) flow and turbine efficiency, using both simulations (Troldborg et al., 2022; Zengler et al., 2024; Revaz and Porté-Agel, 2024) and wind tunnel experiments (Dar et al., 2023).



560 **Figure 18: Individual turbine efficiency ( $C_p$ ) vs. freestream pressure coefficient. The pressures are extracted from the freestream simulation at the turbine location and 3 D downstream at hub height at  $y = 0$  m. Power is extracted from isolated turbine simulations. Cases without buoyancy are marked with two asterisks in the legend.**



The factors influencing  $C_P$  in this study do not appear to be limited to terrain-induced streamwise pressure variations. Within narrow bands of freestream pressure coefficient there are significant differences in  $C_P$  in Fig. 18. Across the flat cases, where pressure coefficient deviates little from zero, turbine efficiency decreases as  $z_i$  decreases, though not as dramatically as the wind farm efficiency in Fig. 16. Stratification above the boundary layer appears, then, to affect turbine blockage and contribute to the  $C_P$  trend in the flat cases and probably other cases as well. There are likely additional contributors: Dar et al. (2023) and Revaz and Porté-Agel (2024) also show clear but imperfect correlations between  $C_P$  and streamwise pressure variation, but in these studies, the other contributors to  $C_P$  variation cannot be related to stable stratification, as the flows are purely neutral, with no buoyancy.

### 3.4 A real wind farm

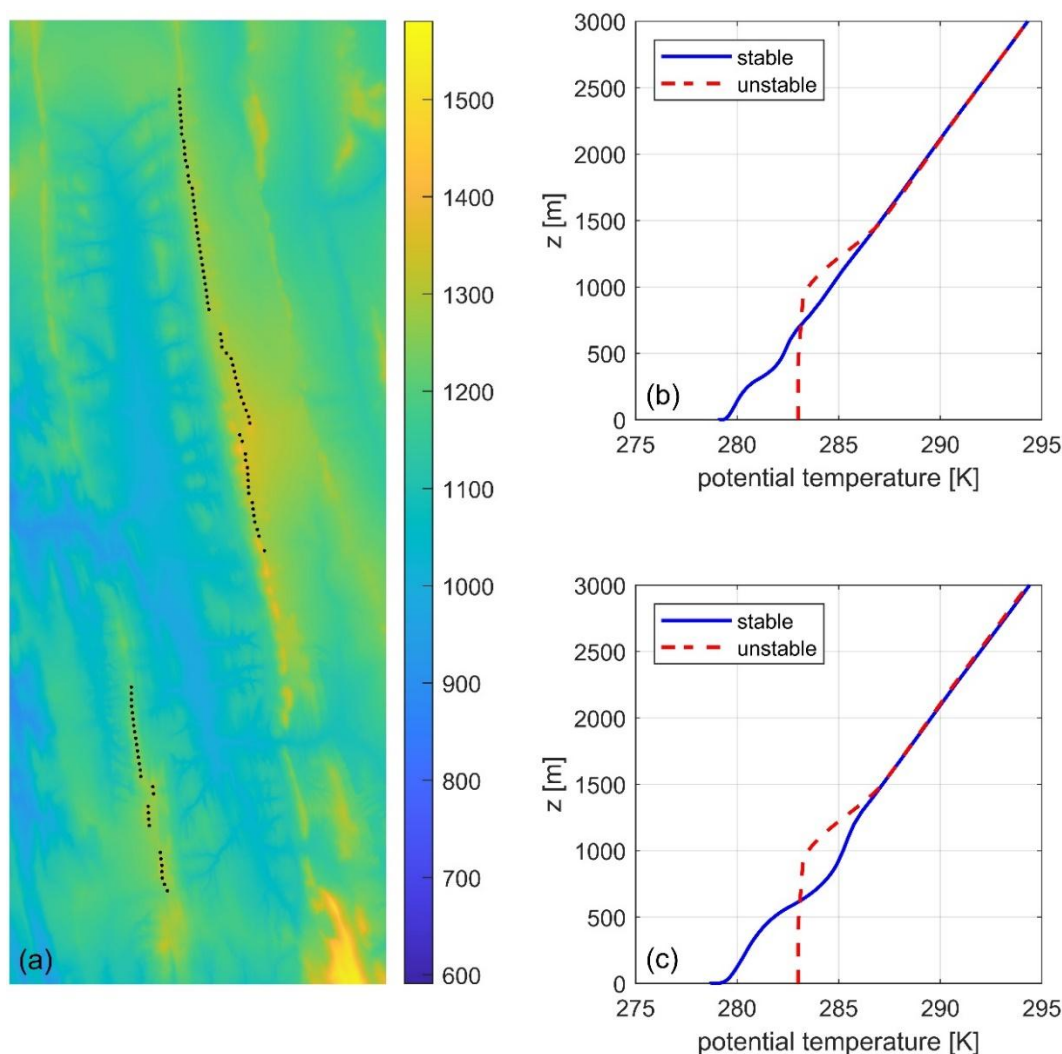
The results presented so far have been from simulations of notional wind farms in conventionally neutral boundary layers on idealized terrain. The results in this subsection are from simulations of a real wind farm with inflow atmospheric conditions derived from the output of a mesoscale model. The subject wind farm has 94 modern wind turbines organized in rows atop ridgelines as depicted in the elevation map in Fig. 19. The spacing between turbines is tight with hub-to-hub distances as low as 1.5 D. The north, south, east, and west domain boundaries are at least 16 km from the turbines, and the top boundary is 17 km above the maximum ground elevation in the simulation.

Results from the Weather Research and Forecasting model (WRF) run with the MYJ boundary layer scheme over a full year were used to inform the RANS inflow boundary conditions. We started with a time series of vertical profiles extracted from the WRF output at a location approximately coincident with the east boundary of the RANS model. We then filtered the time series of profiles for eastern flows and divided the remaining records into two batches, one corresponding to unstable surface conditions and the other corresponding to stable surface conditions. For each batch of records, we derived a single set of profiles (potential temperature, velocity, and turbulence quantities) intended to represent the average conditions in the batch. The inflow potential temperature profiles for the RANS simulations are shown in Fig. 19b and Fig. 19c. The profiles in Fig. 19b were used in the simulations of the wind farm on real terrain. The profiles in Fig. 19c were used in the simulations of the wind farm layout on flat terrain. The inflow wind speed levels were set such that the turbines across all the cases experience similar wind speeds. This can be done while maintaining the same potential temperature profiles for the flat and real terrain cases when the simulated boundary layer is unstable; however, in our approach to simulating stable boundary layers (Bleeg et al., 2015a), the differences in the inflow wind speeds between the flat and real terrain cases necessitate somewhat different potential temperature profiles.

The unstable potential temperature profile does not include a superadiabatic lapse rate in the surface layer. We do this to avoid the occurrence of thermal convections in the solution, as convective structures resolved by a RANS model would likely be spurious. We instead simulate the unstable boundary layer as a conventionally neutral boundary layer but with



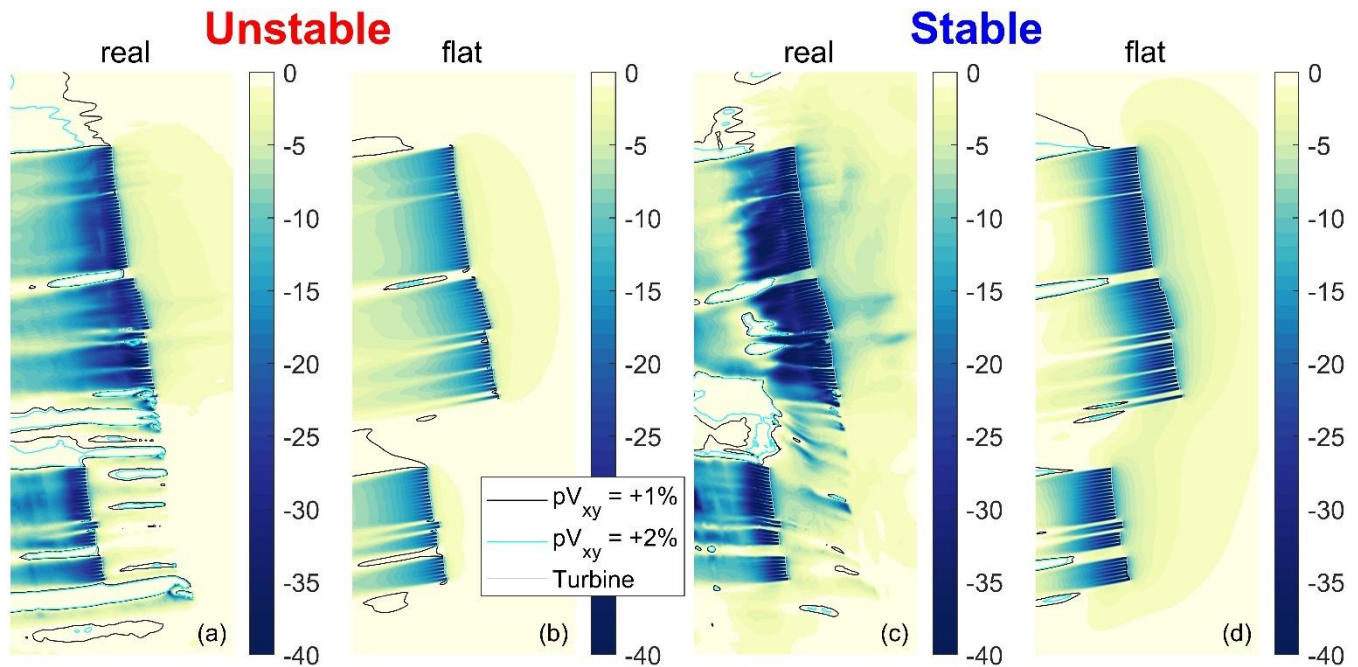
- 595 additional turbulence source terms designed to represent the impact of convections on turbulence, and in turn the mean flow. The source terms are based on similarity theory (Vulfson and Nikolaev, 2022) with the main inputs being ground heat flux and the height of the boundary layer ( $158 \text{ W/m}^2$  and  $1000 \text{ m}$  in this case). Bleeg et al. (2023) has more information on this approach to simulating unstable boundary layers in RANS.
- 600 The average freestream hub-height wind speed across the turbines in all these simulations is approximately  $8 \text{ m/s}$ . The average wind direction is  $80^\circ$ . Figure 20 depicts the results from these simulations.



605 **Figure 19: Elevation map and turbine layout (a); inflow potential temperature profiles for the real terrain simulations (b); and inflow potential temperature profiles for the flat terrain simulations (c).**



In the real terrain results (Fig. 20a and Fig. 20c) the wakes downstream of the wind farm are generally deeper than the wakes in the flat terrain results (Fig. 20b and Fig. 20d). The blockage-related upstream slowdowns in real terrain are also more pronounced, with the slowdowns greater in the simulations of the stable boundary layer than the unstable boundary layer for both real and flat terrain. The loss trends between the cases reflect the difference in upstream blockage effects. In both  
 610 unstable and stable conditions, the turbine interaction loss is about 4% greater in real terrain than in flat terrain for the simulated conditions. In unstable conditions the predicted loss is 5.3% in real terrain and 0.9% in flat terrain. In stable conditions the loss is 16.5% in real terrain and 12.4% in flat terrain. Note that the loss predictions in this paper correspond to the simulated conditions only; if the turbine interaction analysis considered the full range of atmospheric conditions at this wind farm, particularly the wind speed distribution, the predicted blockage-related losses would be much lower.



**Figure 20: Top view of the percent change in wind speed relative to freestream at hub height for the simulation of the unstable boundary layer in real terrain (a); the unstable boundary layer in flat terrain (b); the stable boundary layer in real terrain (c); and the stable boundary layer in flat terrain (d).**

It is less straightforward to assess the effect of terrain on turbine efficiency in this case compared with the idealized cases. This is due to the complexity of the terrain in combination with the fact that, unlike the  $C_p$  curve in Fig. 2, the  $C_p$  curve of a real wind turbine is not constant over a wide range of wind speeds. Nevertheless, we were able to assess the predicted impact of terrain on  $C_p$  for a few of the turbines at this wind farm—two turbines during unstable conditions and four turbines during stable conditions. When the simulated boundary layer was unstable, terrain reduced the average  $C_p$  of the two turbines by  
 625 4.4%. When the simulated boundary layer was stable, terrain reduced the average  $C_p$  of the four turbines by 4.8%. We have not assessed the impact of terrain on  $C_p$  for the other 90+ wind turbines or for other wind directions. Also, as with wind farm



efficiency, the average effect of terrain on turbine efficiency will be much lower when aggregated over the full range of wind speeds.

630 In this subsection simulations of a real wind farm with characteristics similar to the idealized wind farms in Section 3.2 – but using site-specific inflow conditions instead of conventionally neutral boundary layers – yield results consistent with the findings derived from the idealized wind farms with respect to the effect of terrain on wakes, blockage, and efficiency. It is not known from these results the extent to which the findings apply to onshore wind farms in general. To understand this better, future work could involve simulating other types of terrain and wind farm layouts.

## 635 4 Discussion

Terrain influences wake and blockage effects at wind-farm-scale and at turbine-scale according to the RANS results presented in Section 3. In this section we discuss the practical implications of these findings, as well as how much weight one should place on the findings given limitations in the modeling and the scope of the investigation. We also discuss the types of field observations that could improve our understanding of the effect of terrain on wakes and blockage.

### 640 4.1 Practical implications

The influence of terrain on wakes and blockage may have significant implications for wind farm energy yield. According to the results of this RANS-based study, terrain can influence not only array efficiency ( $\eta$ ), but also turbine efficiency ( $C_P$ ). Further, there might be a correlation between the effect of terrain on  $\eta$  and the effect of terrain on  $C_P$ . In the single-row wind farms in this study, the correlation coefficient between  $\eta$  and  $C_P$  is 0.79.

645 Take for example Case C, a single row on the leeward side of the hill with  $z_i = 960$  m. In this case, the turbines collectively produce 9.31% less power than they would if each one were simulated in isolation. By comparison, the turbine interaction loss,  $L$ , in the corresponding flat terrain analysis, Case D, is 1.46%. Thus, the wind farm turbines in Case C are 7.85% less efficient relative to isolated operation than the turbines in Case D. On top of that, the isolated turbines in Case C are also less efficient, with a  $C_P$  of 0.421 compared to 0.444 in Case D, a difference of 5.81%.

655 Current industry practices for estimating the energy yield of a planned wind farm largely neglect the effect of terrain on wakes and blockage, as well as the related effect terrain has on wind farm efficiency and individual turbine efficiency. This could lead to material prediction errors, which would contribute to uncertainty in energy yield predictions. These errors may also result in wind farms that are less productive than they otherwise would be if terrain effects were more fully accounted for during the design of the wind farm.



Mitigating this risk requires flow modeling that captures the key fluid dynamic elements contributing to the influence of terrain on wakes and blockage. Specifically, terrain-induced streamwise pressure gradients can have a large impact on wake development, which in turn influences upstream blockage effects. The inviscid response of stratified flow to the terrain and wind farm further modifies blockage effects. Completing the interlinkage, blockage effects influence wake development (Lanzilao and Meyers, 2024; Ndindayino et al., 2025). In other words, terrain affects wakes, which affect blockage; terrain also directly influences blockage, which affects wakes. It is a tightly coupled fluid dynamic system, and the approach used to model it should aim to reflect this overarching aspect of the flow.

## 4.2 Limitations

How seriously one should take these conclusions depends in no small part on the reliability and the representativeness of the simulations behind them. With respect to real-world representativeness, the effect of terrain on wakes and blockage may be generally larger in idealized quasi-2D terrain than in real terrain. The effect may also be larger, or at least different, for wind farms organized in rows with tight spacing as compared to other types of turbine layouts. The simulated atmospheric conditions also raise questions of representativeness: the effect of terrain on stratified flow is very sensitive to the simulated atmospheric conditions, particularly the potential temperature profile; if the atmospheric conditions simulated in an analysis of a particular wind farm are not representative of the actual site conditions, it could result in substantial errors in the prediction of the impact of terrain on wake and blockage effects.

With respect to reliability of the simulations, the RANS model used in this study has shown good agreement with field observations related to blockage and wakes at many wind farms, including three with elevation variation on the order of what is simulated in this study (these three are internal validations not reported in the references provided in Section 2.1). That said, there are reasons to question the near-wake predictions. Simulations with actuator disks cannot capture some of the important flow features in a near wake, like hub and tip vortices, and internal validation indicates that the near wakes simulated with this RANS model generally diffuse too rapidly relative to field observations. The effect of terrain-induced streamwise pressure gradients on the near wake of an individual turbine is proposed to be central to the impact of terrain on turbine efficiency. Given these facts, there is enhanced risk that the predicted turbine efficiency trends are inconsistent with what would occur in the field. That said, several other recent studies cited herein, including a wind tunnel experiment, have yielded similar trends.

## 4.3 Field observations wanted

Relevant field observations could improve our understanding of the impact of terrain on wakes and blockage as well as our understanding of the reliability of related flow modeling. Wind speed measurements taken just upstream of a wind farm for sufficiently long periods before and after the wind farm starts operating could help in assessing the effect of terrain on wind farm blockage. As described in Bleeg et al. (2018), an analysis of the data can reveal the degree to which the wind speed



690 upstream of the wind farm changes relative to a reference wind speed time series due to the presence of the wind farm.  
Separately, turbine power data from a wind farm located in terrain where there are marked streamwise variation in the  
background wind speed could help in the assessment of the effect of terrain on wakes.

However, it is difficult to isolate the impact of terrain on field observations related to blockage and wakes. Unlike in a wind  
695 tunnel or a simulation, it is not reasonably possible in the field to obtain back-to-back measurements of a similar wind farm  
in flat terrain. A way forward could be to compare the measurements with simulations of the wind farm layout on the real  
terrain and on flat terrain. The comparison could provide insight into both the reliability of the simulations and the degree to  
which terrain is affecting blockage and wake effects at the subject wind farm.

700 As for the effect of terrain on turbine efficiency, a type of measurement commonly used in the wind industry may be  
relevant. The theory promoted here, and elsewhere, is that terrain can affect induction and in turn  $C_p$ . This terrain effect, if  
real, is likely reflected in standard turbine power performance measurements (PPM), which involve concurrent  
measurements of turbine power and upstream wind speed. The IEC standard for turbine PPM, IEC 61400-12 (2022), requires  
that the wind speed measurement take place 2 to 4 rotor diameters upstream of the test turbine, where the effect of turbine  
705 induction is small (Medici et al., 2011). Thus, any impact of terrain on induction should have less of an impact on the  
measured wind speed than on the wind speed at the rotor face. Therefore, if terrain has an impact on  $C_p$ , it should be at least  
partly captured in the PPM. That said, even if the effect of terrain is captured in the measurements, it will not be  
straightforward to isolate the effect. An assessment of many measured  $C_p$  curves from a single turbine model across a variety  
of topographies may help – especially if done in conjunction with simulations with and without the real terrain. Still, there  
710 are many confounding factors that can influence measured power performance—for example, flow inclination and shear,  
both of which are sensitive to terrain—and it may be difficult to truly isolate from the measurements the effect of terrain on  
induction and in turn turbine efficiency.

## 5 Conclusions

We simulated 35 different combinations of wind farm layout, terrain, and inflow atmospheric conditions using RANS. The  
715 wind farms were organized in one or two rows of closely spaced turbines. The terrain in nearly all cases was idealized, either  
quasi-2D or flat, and the inflow boundary layers in these cases were neutral with overlying thermal stratification. The few  
remaining cases correspond to a real wind farm with characteristics similar to some of the idealized wind farms but  
simulated in a stable or an unstable boundary layer. Terrain had a pronounced influence on wake and blockage effects in the  
simulation results. Further, the RANS-predicted wind farm efficiencies and turbine efficiencies in terrain, whether idealized  
720 or real, departed markedly from back-to-back RANS-predictions on flat ground.



Hill-induced streamwise pressure gradients can have a large influence on wake development, and we propose, based on the results of this study, that this effect can feed back to influence upstream slowdowns related to blockage. Further analysis of the results, including an assessment of local Froude number, suggests that the inviscid response of stratified flow to a wind farm in terrain can additionally affect wind farm blockage – in a way related to but distinct from its effect on blockage via terrain-induced pressure gradients as they pertain to wake recovery.

This work differs in a few ways from earlier studies on the combination of terrain and wind turbines. Perhaps most notably, we consider the effect of terrain on both wakes *and* blockage. And whereas previous studies were limited to purely neutral flows, with no buoyancy, this study includes the effects of thermal stratification, which can significantly modify flow over terrain and its response to the presence of a wind farm. Finally, by virtue of simulating full wind farms, we directly connect the impact of terrain to wind farm efficiency.

That said, this study is not without its own limitations. From the terrain to the wind farm layouts to the atmosphere, the simulated conditions are mostly idealized. The simulations of an existing wind farm mitigate this concern somewhat in that the results are reasonably consistent with the findings from the idealized wind farms. Nevertheless, the general applicability of some of these findings can be fairly questioned due to the limited range of layout types and terrain types investigated. An analysis of a broader range of wind farms could help provide a more comprehensive understanding of the impact of terrain on wakes and blockage. Further, relevant, publicly available field observations could help improve confidence in the model and the study findings.

In the meantime, between this study and those that came before, there appears to be enough evidence to make the following claim: At proposed wind farm sites where terrain induces significant streamwise variation in wind speed, energy yield prediction approaches that do not capture the effect of terrain on wakes and blockage are at increased risk of elevated prediction errors. This could materially increase uncertainty in the energy yield predictions and perhaps even lead to less efficient wind farm layout designs.

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## References

- Allaerts, D. and Meyers, J. Sensitivity and feedback of wind-farm-induced gravity waves, *J. Fluid Mech.* 862, 990-2028, doi:10.1017/jfm.2018.969, 2019.
- Baines, P.G. *Topographic Effects in Stratified Flow*, Second Edition, Cambridge University Press, Cambridge, UK, 2022.
- 755 Bayron, P., Kelso, R., Chin, R. Experimental study on the evolution of wind turbine wake under streamwise pressure gradients, *Energy Conversion and Management: X*, 20, 100479, doi: 10.1016/j.ecmx.2023.100479, 2023.
- Bleeg, J., Digraskar, D., Woodcock, J., Corbett, J.F. Modeling stable thermal stratification and its impact on wind flow over topography, *Wind Energy*, 18, 369-383, doi:[10.1002/we.1692](https://doi.org/10.1002/we.1692), 2015.
- Bleeg, J., Digraskar, D., Horn, U., Corbett, J.F. Modelling Stability at Microscale, Both Within and Above the Atmospheric  
 760 Boundary Layer, Substantially Improves Wind Speed Predictions, *Proceedings of the EWEA Conference*, Paris, France, 17-20 November, 2015.
- Bleeg, J., Purcell, M., Ruisi, R., Traiger, E. Wind Farm Blockage and the Consequences of Neglecting Its Impact on Energy Production, *Energies*, 11, 1609, doi:10.3390/en11061609, 2018.
- Bleeg, J. and Montavon, C. Blockage effects in a single row of wind turbines, *J. Phys. Conf. Ser.*, 2265, 022011, doi:  
 765 10.1088/1742-6596/2265/2/022001, 2022.
- Bleeg, J., Del Hoyo, M., Montavon, C. Modelling the convective boundary layer at microscale using RANS, *WESC*, Glasgow, UK, 23-26 May, 2023, <https://doi.org/10.5281/zenodo.15039844>, 2023.
- Bleeg, J., Vishwakarma, P., Del Hoyo, M., Simmons, L. The impact of blockage and wakes on seven power performance tests conducted at two wind farms, *J. Phys. Conf. Ser.*, 2767, 042026, doi:10.1088/1742-6596/2767/4/042026, 2024.
- 770 Bortolotti, P., Tarres, H.C., Dykes, K., Merz, K., Sethura, L. IEA Wind Task 37 on Systems Engineering in Wind Energy – WP21 Reference Wind Turbines, International Energy Agency, NREL/TP-73492, 2019.
- Cai, T., Cheng, S., Segalini, A., Chamorro, L. Local topography-induced pressure gradient effects on the wake and power output of a model wind turbine, *Theor. Appl. Mech. Lett.*, 11, 100297, doi:10.1016/j.taml.2021.100297, 2021.
- Dar, A.S. and Porté-Agel, F. An Analytical Model for Wind Turbine Wakes under Pressure Gradient, *Energies*, 15, 5345, doi:10.3390/en15155345, 2022.
- 775 Dar, A.S., Gertler, A.S., Porté-Agel, F. An experimental and analytical study of wind turbine wakes under pressure gradient, *Phys. Fluids*, 35, 045140, doi: 10.1063/5.0145043, 2023.
- Hyvärinen, A., Lacagnina, G., Segalini, A. A wind-tunnel study of the wake development behind wind turbines over sinusoidal hills. *Wind Energy*, 21: 605-617, doi:10.1002/we.2181, 2018.
- 780 International Electrotechnical Commission: Power performance measurements of electricity producing wind turbines, IEC 61400-12-1:2022, <https://webstore.iec.ch/en/publication/68499>, 2022.
- Kashyap, P.P.K.: *Impact Assessment of Wind Farm Blockage in Complex Terrain*. M.S. thesis, DTU, 2022.



- Lanzilao, L. and Meyers, J. A parametric large-eddy simulation study of wind-farm blockage and gravity waves in conventionally neutral boundary layers. *J. Fluid Mech.*, 979, A54, doi:10.1017/jfm.2023.1088, 2024.
- 785 Liu, L. and Stevens, J.A.M.: Effects of Two-Dimensional Steep Hills on the Performance of Wind Turbines and Wind Farms, *Boundary-Layer Meteorol.*, 176, 251-269, doi:10.1007/s10546-020-00522-z, 2020.
- Medici, D., Ivanell, S., Dahlberg, J.A., Alfredsson, P.H. The upstream flow of a wind turbine: blockage effect, *Wind Energy*, 14, 691-697. <https://doi.org/10.1002/we.451>, 2011.
- Montavon, C., Blegg, J., Riechert, J., Steger, M., Soderberg, S., Schmitt, C. Measuring and Modelling Wind Farm Blockage  
 790 Offshore, Wind Europe Technology Workshop, Online, 8-10 September, 2021.
- Montavon, C., Steger, M., Blegg, J., Del Hoyo, M., Menke, R., Schmitt, C., Riechert, J., Rautenstrauch, J. Blockage and cluster-to-cluster interactions from dual scanning lidar measurements, WESC, Glasgow, UK, 23-26 May, 2023, <https://doi.org/10.5281/zenodo.8000511>, 2023.
- Montavon, C., Rodaway, C., Gunn, K., Smith, G., Dunsmore, D., Del Hoyo, M., Sinclair, K. Cluster wakes and their effect  
 795 on a wind farm annual energy production, White Paper, DNV and RWE, 2024.
- Ndindayino, O., Puel, A., and Meyers, J. Effect of blockage on wind turbine power and wake development, *Wind Energy Science*, 10, 2079-2098, <https://doi.org/10.5194/wes-10-2079-2025>, 2025.
- Patel, J.A., Maity, A., Ghaisas, N.S. The Influence of Topographic Variations on Wind Turbine Wake Characteristics Using LES, *J. Phys. Conf. Ser.*, 2767, 092086, doi:10.1088/1742-6596/2767/9/092086, 2024.
- 800 Politis, E.S., Prospathopoulos, J., Cabezon, D., Hansen, K.S., Chaviaropoulos, P.K., Barthelmie, R.J.: Modeling wake effects in large wind farms in complex terrain: the problem, the methods, and the issues, *Wind Energy*, 15, 161-182, [doi:10.1002/we.481](https://doi.org/10.1002/we.481), 2012.
- Radünz, W.C., Carmo, B., Lundquist, J.K., Letizia, S., Abraham, A., Wise, A., Gomez, M.S., Hamilton, N., Rai, R.K., Peixoto, P.S. Influence of simple terrain on the spatial variability of a low-level jet and wind farm performance in the  
 805 AWAKEN field campaign, *Wind Energy Science*, 10, 2365-2393, doi:10.5194/wes-10-2365-2025, 2025.
- Revaz, T. and Porté-Agel, F. Effect of hills on wind turbine flow and power efficiency: A large-eddy simulation study, *Phys. Fluids*, 36, 095180, doi:10.1063/5.0226544, 2024.
- Shamsoddin, S. and Porté-Agel, F. Wind turbine wakes over hills, *J. Fluid Mech.*, 855, 671-702, doi:10.1017/jfm.2018.653, 2018.
- 810 Siemens PLM Software: User Guide for STAR-CCM+, Version 2022.1; Plano, TX, USA, 2022.
- Tian, W., Ozbay, A., Yuan, W., Sarakar, P., Hu, H. An Experimental Study on the Performances of Wind Turbines over Complex Terrain, 51<sup>st</sup> AIAA Aerospace Sciences Meeting, Grapevine, TX, 7-10 January, doi:10.2514/6.2013-612, 2013.
- Troldborg, N., Andersen, S.J., Hodgson, E.L., Meyer Forsting, A. Brief communication: How does complex terrain change  
 815 the power curve of a wind turbine? *Wind Energy Science*, 7, 1527-1532, doi:10.5194/wes-7-1527-2022, 2022



- van der Laan, M., Sørensen, N.N., Réthoré, P.E., Mann, J., Kelly, M.C. The  $k$ - $\varepsilon$ - $f_p$  model applied to double wind turbine wakes using different actuator disk force methods, *Wind Energy*, 18, 2223-2240, doi:10.1002/we.1816, 2014.
- Vosper, S.B. Inversion effects on mountain lee waves, *Q.J.R. Meteorol. Soc.*, 130, 1723-1748, doi:[10.1256/qj.03.63](https://doi.org/10.1256/qj.03.63), 2004.
- 820 Vulfson, A. and Nikolaev, P. Local similarity theory of convective turbulent layer using “spectral” Prandtl mixing length and second moment of vertical velocity, *J. Atmos. Sci.*, 79, 101-118, doi:[10.1175/JAS-D-20-0330.1](https://doi.org/10.1175/JAS-D-20-0330.1), 2022.
- Zengler, C.P., Troldborg, N., Gaunaa, M. Is the free wind speed sufficient to determine aerodynamic turbine performance in complex terrain? , *J. Phys, Conf. Ser.*, 2767, 092049, doi:10.1088/1742-6596/2767/9/092049, 2024.
- Zhang, M. Du, B., Ge, M., Liu, Y. Influence of pressure gradient on the wake evolution of aligned wind turbines: A large-eddy simulation study, *IET Conference Proceedings*, 15, 1240-1244, doi: [10.1049/icp.2023.2479](https://doi.org/10.1049/icp.2023.2479), 2023.
- 825