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### Case study of a bore wind-ramp event from lidar measurements and HRRR simulations over ARM Southern Great Plains Yelena L. Pichugina<sup>1, 2</sup>, Robert M. Banta<sup>1, 2</sup>, E.J. Strobach<sup>1, 2</sup>, B.J. Carroll<sup>1, 2</sup>, W. Alan Brewer<sup>2</sup>, 3 D.D. Turner<sup>3</sup>, V. Wulfmeyer<sup>4</sup>, E. James<sup>3</sup>, T.R. Lee<sup>5</sup>, S. Baidar<sup>1,2</sup>, 4 J.B. Olson<sup>3</sup>, R.K. Newsom<sup>6</sup>, H.-S. Bauer<sup>4</sup>, and R. Rai<sup>6</sup> 5 6 <sup>1</sup>CIRES, University of Colorado, Boulder, CO 80309, USA <sup>2</sup>NOAAChemical Sciences Laboratory, Boulder, CO 80305, USA <sup>3</sup>NOAA Global Systems Laboratory, Boulder, CO 80305 USA <sup>4</sup>University of Hohenheim, Institute of Physics and Meteorology, Stuttgart, 70599, Germany 10 <sup>5</sup>NOAA Air Resources Laboratory, Oak Ridge, TN 37830, USA <sup>6</sup>Pacific Northwest National Laboratory, Richland, WA 99352, USA Correspondence to: Yelena Pichugina (Yelena.Pichugina@noaa.gov) Abstract The rapid change of wind speed and direction on 21 August 2017 is studied using Doppler lidar measurements at 5 sites of the Atmospheric Radiation Measurement (ARM) Southern Great Plains (SGP) facility in north-central Oklahoma. The Doppler lidar data were investigated along with meteorological variables such as temperature, humidity, and turbulence available from the large suite of instrumentation deployed at the SGP Central Facility (C1) during the Land-Atmosphere Feedback Experiment (LAFE) in August 2017. Lidar measurements at 5 sites, separated by 55 - 70 km, allowed us to document the development and evolution of the wind flow over the SGP area, examine synoptic conditions to understand the mechanism that leads to the ramp event and estimate the ability of the High-Resolution Rapid Refresh (HRRRv3) model to reproduce this event. The flow feature in question is an atmospheric bore, a small-scale phenomenon that is challenging to represent in models, that was generated by a thunderstorm outflow northwest of the ARM SGP area. The small-scale nature of bores, its impact on power generation, and the modeling challenges associated with representing bores are discussed in this paper. The results also provide information about model errors between sites of different surface and vegetation types.

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In wind energy, rapid changes in wind speed over a short period of time ranging from a

few minutes up to several hours are defined as wind "ramp events". These wind ramps may lead

to significant fluctuations in the power generated by wind turbines. In practice, an increase in wind

speed or wind power has been defined as a "ramp-up", and a decrease, as a "ramp-down" event

(Lee et al. 2012, Worsnop et al. 2018). Fluctuations of wind-turbine power on sub-hourly, hourly,

or daily time scales may affect the overall power generated by the wind plant and may bring large

relatively short time, but various definitions of power ramps are provided in the literature,

depending on the magnitude and duration of the event, where magnitude is considered with respect

to the rated power (P<sub>rated</sub>) of the wind farm. The limits on key quantities that define ramps – the

change in wind power production ( $\Delta P$ ) and the period ( $\Delta t$ ) of this change vary among studies. For

example, Gallego et al. (2014) defined power ramps as having  $\Delta t$  ranging from 30 min to 3 hours

and  $\Delta P$  within 20 -75% of the turbine-rated power; Greaves et al. (2009) defined power ramps

with  $\Delta P$  about 50% of the installed wind capacity that occurs within less than 4 hours; Bossavy et

al. (2010) define a power ramp as when a  $\Delta P$  of ~ 50% occurs over one hour. In addition to the

magnitude  $\Delta P$  and duration  $\Delta t$ , power ramps have been characterized in the literature by the ramp

Wind-power ramps have been broadly defined as significant changes in production over a

uncertainties to power scheduling and trading (Ela and Kemper, 2009).

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1. Introduction

49 rate  $\Delta P / \Delta t$ , the starting or central time of the event, the type of the ramp (ramp-up or ramp-down), 50 and the ramp gradient (Sherry and Rival, 2015; Ferreira et al. 2010). Other ramp features, such as 51 ramp shape, diurnal cycle, and seasonality, are discussed in several case studies (Pichault et al., 52 2020). 53 Power ramps associated with a significant increase or decrease of wind speed (wind ramps) 54 can be driven by different atmospheric conditions (Freedman et al. 2008). At local scales, 55 processes driven by horizontal gradients of surface heating, such as sea breezes (Wharton et al. 56 2011), sea-breeze-generated marine intrusions (Banta et al. 2020), and cold-pools (Pichugina et al. 57 2019; Wilczak et al. 2019; McCaffrey et al. 2019; Adler et al. 2023) can cause wind ramps. The

58 atmospheric flow phenomena known as low-level jets (LLJs), can also amplify wind speed 59 (Freedman et al. 2008). LLJs have been studied for their meteorological importance and frequency, 60 especially in the Great Plains where they are present on about 20% to 65% of days depending on 61 the season and year (Bonner 1968; Mitchell et al. 1995; Whiteman et al. 1997; Song et al. 2005; 62 Carroll et al. 2019). Along with the direct importance for wind energy through strong wind speed 63 and shear, LLJs can indirectly impact productivity by promoting convection through moisture 64 transport and low-level convergence (Geerts et al. 2015; Berg et al 2015). Basic information on 65 LLJs is provided in Appendix B. More information on LLJ properties and nighttime evolution 66 analyzed from Doppler lidar measurements as well as the variability between ARM SGP sites can 67 be found in Pichugina et al. (2023).

The presence of LLJs and the post-sunrise growth of the planetary boundary layer have been identified as dominant factors of ramp events in Deppe et al. (2012). Shorter-duration power ramps are mainly influenced by mesoscale systems, whereas synoptic systems tend to be responsible for longer-duration wind and power ramps (Drew et al. 2018). At larger scales, features such as frontal passages, density currents, and thunderstorm outflows (Freedman and Zack, 2012; DeMarco and Basu, 2018) can lead to significant changes in the wind flows.

74 Accurate model prediction of ramp events is necessary to anticipate and mitigate negative 75 effects on wind-energy resource management. The improvement of models used in the wind-76 energy industry, from frequent updates of model physics, parameterization schemes, and 77 horizontal grid spacing, requires evaluation of model skills through comparisons with observations 78 (Olson et al. 2019, Shaw et al. 2019, Wilczak et al. 2019, Banta et al. 2023). Dedicated field 79 campaigns have been conducted to address this challenge. The High-Resolution Rapid Refresh 80 (HRRR) numerical weather prediction model was continuously updated during the second Wind 81 Forecast Improvement Project (WFIP2, Olson et al. 2019) and all improvements from version 1 82 (HRRRv1) to version 4 (HRRRv4) were evaluated against various types of remote sensors (lidars, 83 sodars, and wind-profiling radars) and in-situ measurements (Banta et al. 2020, 2021, 2023; Bianco 84 et al. 2019; Draxl et al. 2021, Olson et al. 2019, Pichugina et al. 2019, 2020, 2022; Rai et al. 2020, 85 Wilczak et al. 2019). Different methods of probabilistic ramp forecasts from the HRRR were 86 compared to measurements from two tall towers located in western Colorado and eastern Oregon 87 of the United States (Worsnop et al 2018), and valuable information was obtained on the 88 uncertainty and improved model skill over the raw forecasts.

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Measurements from remote-sensing instruments distributed over an area may provide insight into wind and power ramps over wind-farms. For example, during WFIP2, fluctuations of wind-speed from scanning Doppler lidar measurements at three sites along the Columbia River Valley approximated the fluctuations of total power generated within the Bonneville Power Authority (BPA) balancing area (Pichugina et al. 2020, Wilczak et al. 2019). The largest power ramps (up to 3 GW or more) were found for westerly gap-flows in summer and cold pools in winter months (McCaffrey et al. 2019, Pichugina et al. 2019, 2020).

96 During the Land-Atmosphere Feedback Experiment (LAFE), conducted at the Central 97 Facility (C1) of the Atmospheric Radiation Measurement (ARM) Southern Great Plains (SGP) 98 atmospheric observatory in August 2017 (Wulfmeyer et al. 2018), fluctuations of nocturnal wind 99 speed were measured by Doppler lidar on several nights. The north-central Oklahoma location of 100 this observatory is known for its nocturnal maximum in thunderstorm activity and precipitation 101 (Wallace 1975; Fritsch et al. 1986; Tripoli and Cotton 1989), and most of the significant wind-102 ramping fluctuations observed were associated with flow features ultimately caused by 103 thunderstorms. The largest ( $\sim 12 \text{ m s}^{-1}$ ) ramp-down was a transient disturbance observed on 21 August 2017 (Fig. 1), starting just before 1100 UTC (UTC = local time + 6h) at site C1. 104 105 Measurements from scanning Doppler lidars were also available at four SGP extended facilities. 106 Datasets from the network of lidars provide an exceptional opportunity to estimate the spatial 107 variability of this significant ramp event over the SGP area, to identify weather-related causes of 108 the sudden change of wind speed, and to quantify the ability of the HRRR NWP-model simulations 109 to capture the ramp event as observed at the five ARM SGP sites on this day.

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111 Figure 1. (a) Time-series of wind speed at 6 heights from lidar measurements at SGP site C1. (b) Wind 112 power calculated for the "virtual" turbine with the rotor diameter of 70 m and wind speed from (a) taken at 113 three heights (74 m, 95 m, and 117 m). Black arrows indicate the magnitude and duration of the power 114 ramp event for the wind turbine with the hub-height of 117 m.

115 The paper is organized as follows. Section 2 provides the location of Doppler lidar sites, a 116 description of available measurements and temporal and vertical resolutions of obtained profiles. 117 A brief overview of the HRRR version 3 (HRRRv3) used in this study is also provided in this 118 section. Section 3 discusses a larger-scale context of the synoptic situation during the ramp-down 119 event on 21 August. Section 4 presents the results of the ramp events from Doppler and Raman 120 lidar measurements at the central facility. Section 5 shows the spatial variability of the wind speed 121 ramp from lidar and measurements at 5 SGP sites, illustrates the influence of the wind ramp on the 122 power production of a hypothetical wind turbine, and provides analysis of winds and AERI-123 measured temperature fields. Section 6 estimates the ability of 3-km HRRRv3 hourly simulations

124 to capture the vertical structure and temporal evolution of the wind ramp event.



125 Temperature and humidity from NOAA/NCRP's North American Regional Reanalysis 126 (NARR) and wind speed from the Experimental High Resolution Rapid Refresh (HRRRX) model 127 are provided in Appendix A to support the possible cause of the observed ramp event.

### 128 **2. Description of SGP lidar sites and measurements.**

129 Scanning Doppler lidars were located at the ARM SGP central facility (C1) and the four 130 extended facilities E32, E37, E39, and E41 separated by 56-57 km along the south-north and 66-131 77 km along the east-west directions (Fig. 2). The sites have different surface and vegetation types 132 from cropland to grassland and pasture, and site elevations vary between 279 and 379 m ASL 133 (Pichugina et al. 2023, Table 1). Sites C1 and E41 are surrounded by wind farms. The closest wind 134 turbine to the lidar at C1 is located to the south at 3.74 km and from the lidar at E41 to the 135 southwest at 2.52 km. The terrain and trees in some areas add more complexity to measured winds 136 and the uncertainty of lidar data from low-elevation scans. Quality control of the line-of-sight 137 velocities (Newsom and Krishnamurty, 2020), allowed the removal of outliers and provided 138 accurate profile data.





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144 The lidars at all five sites are Halo Photonics Stream Line scanning systems that 145 continuously operated in synchronized scanning modes providing multi-year datasets of wind 146 speed, wind direction, and three components of the wind vector. Profiles of all these variables 147 from 91 m up to several km above ground level (AGL), obtained at a temporal resolution of 10-148 min and a vertical resolution of ~26 m are available at the ARM SGP archive 149 https://www.arm.gov/capabilities/observatories/sgp. Details of the deployment history, raw and processed data description, along with other valuable information and sample plots, can be found 150 151 in Newsom and Krishnamurty (2020).

### 2.1. Instrumentation deployed for LAFE

153 In August 2017, additional measurement systems, including various types of lidar, were 154 deployed to the central facility (LAFE; Wulfmeyer et al. 2018). The German University of 155 Hohenheim (UHOH) deployed a scanning Doppler lidar (ULID) that operated at C1 from 13 Aug 156 to 6 Sep 2017 in a six-beam VAD scanning mode to obtain profiles of wind speed and direction 157 having a temporal resolution of 1 min and turbulence profiles at a temporal resolution of 5 min. 158 Turbulence profiles include TKE, vertical fluxes of horizontal momentum, and variances of wind 159 vector components, computed from a 6-beam measurement technique (Sathe et al. 2015; Bonin et 160 al. 2017). The vertical resolution of all variables is ~21 m. In addition, the UHOH Raman Doppler 161 lidar (URLID, Hammann et al. 2015; Behrendt et al. 2015) provided temperature and humidity 162 profiles at 10-min temporal and 30-m resolutions from 30 m up to several km AGL. During the 163 LAFE, the ARM SGP lidar (SLID) at C1 operated in staring mode providing only vertical-velocity 164 data. All three lidars (ULID, SLID, and URLID) were deployed side-by-side (Wulfmeyer et al. 165 2018).

Various data, including wind and temperature, were also available at C1 from sonic anemometers installed at 25 m and 60 m on the 80-m meteorological tower located ~ 250 m from SLID. Standard meteorology, surface fluxes, soil temperature, moisture, and radiation were also available at 2.5 and 10 m from three NOAA / ARL surface energy balance towers (Lee and Buban, 2020; Lee et al. 2021) located along a southwest-northeast transect at distances of ~ 0.7-1 km over cropland (Towers 1 and 3) and the natural-vegetation mix (Tower 2).

172 2.2. HRRRv3: The model used in the study

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173 The NOAA-developed numerical weather prediction (NWP) model (Benjamin et al. 2016) 174 used in this study is an updated third version of the 3-km HRRR (High Resolution Rapid Refresh) 175 model, i.e., HRRRv3. The HRRRv3 domain encompasses the continental United States, providing 176 wind forecasts every hour (Table 1). The HRRRv3 was in the transition-to-operations stage during 177 2016-2018 and became operational in October 2018 (James et al. 2022). Table 1 provides the 178 physics configuration of the HRRRv3 model, adapted from Dowell et al. 2022. The previous 179 version of the HRRR, the HRRRv2, operational during Oct 2016-Oct 2018 (James et al, 2022), is 180 used in some examples to illustrate the difference between HRRR versions. The full description 181 of the HRRRv3 along with other HRRR versions (HRRRv1-HRRRv3), modifications, and 182 improvements can be found in Dowell et al. (2022), and the improvements in performance for 183 specific forecast problems are documented for all HRRR versions in James et al. (2022) along with 184 a comprehensive list of previously published articles evaluating specific aspects of HRRRv3 185 performance.

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Table. 1 Physics configuration for HRRRv3.

Model	WRF-ARWv3.8.1+
Domain	CONUS, Alaska
Init Frequency	1h, 3h
Мар	Lambert conformal projection stereographic (AK)
Grid points (x,y)	1800 x 1060; 1300 x 920
Grid spacing	3 km
Vertical layers	51
Pressure top	20 hPa
Lateral boundary conditions	RAP
Initial conditions	RAP post-DFI plus;
	1-h spinup, 3-km GSI with GDAS
Vertical coordinate	Hybrid sigma - terrain-following
Horizontal/ Vertical advection	Fifth-order upwind
Scalar advection	Positive definite advection
Large time step	20 s
Upper -level damping	Rayleigh, dampcoef = 0.2 s-1, zdamp =5000 m
<b>Computational horizontal</b>	6th order (0.25), horizontal only (not on slopes),
diffusion	applied to all variables
Run frequency	Hourly
Forecast duration	36h every 6h, otherwise 18h
Radiation	RRTMG

Land surface, including	
number of layers	RUC LSM, 9 soil levels, 2-layer snow (v3.8+)
Land use	30" MODIS
Planetary boundary and	
surface layer	Mellor-Yamada- Nakanishi-Niino (v3.8+)
Subgrid-scale clouds	MYNN prognostic SGS cloud fraction, cloud water
Orographic drag	small-scale orographic drag
Cloud Microphysics	Thompson- Eidhammer "aerosol aware" (v3.8)
Microphysics temperature	
tendency limit	0.07 K s <sup>-1</sup>

187 For verification of HRRRv3 against lidar measurements, the gridded model output was 188 extracted at the position of the lidar by bilinear interpolation from the surrounding four grid points. 189 Using other extraction techniques, such as cubical interpolation or taking output from the nearest 190 grid point, show similar results with correlation coefficients of 0.99 between the extraction 191 techniques, differences in mean wind speed of 0.01 to 0.22 m s<sup>-1</sup>, and standard deviations < 0.085192 m s<sup>-1</sup> (Pichugina et al. 2020). Modeled values obtained at the location of each lidar were then 193 linearly interpolated to the heights of lidar measurement. The effects of the vertical-interpolation 194 method and uncertainties of two approaches-first, when measurements are interpolated to model 195 output levels (lidar-to-model), and second, when modeled variables are interpolated to the heights 196 of lidar measurements (model-to-lidar)-are discussed and the second approach (model-to-lidar) 197 is adopted, as justified in Pichugina et al. (2017).

For quantitative comparisons of modeled and measured wind-speed profiles, lidar measurements were hourly averaged to match the time interval of model output.

### 200 3. Larger-scale context

### 201 3.1 Great Plains nocturnal thunderstorms, gust fronts, and bores

Thunderstorms generate cold outflows and density currents (gust fronts). When these outflows push through a surface-based stable layer, such as a nocturnal inversion, they can create a wave ahead of the gust front in the form of a bore or a solitary wave, as shown by Knupp (2006). Bores are a form of gravity wave in the lower atmosphere, representing a superposition or "packet" of gravity wavelengths. They form in and propagate through a layer of positive static stability (i.e., where potential temperature  $\theta$  increases with height *z*, or  $d\theta/dz > 0$ ), and nighttime temperature inversions often provide good conditions for bore propagation. The speed of propagation of the

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waves increases with stronger atmospheric stability. NWP modeling studies suggest that the speed may also depend on the depth of the stable layer ahead of and behind the bore (e.g., Osborne and Lapworth 2017), and that other conditions, such as the alignment or misalignment of forcing associated with the large-scale synoptic and the mesoscale phenomenon that generates bores, can modify convergence and bore propagation. Moreover, the influence of diabatic heating is a relatively unexplored topic that additionally adds to the complexity of gust front/bore evolution.

Bores form when the stable flow encounters an obstacle, which can be moving. The bore 215 216 develops ahead (upwind) of the obstacle. Over the U.S. Great Plains, as described, this obstacle is 217 often a thunderstorm gust front, an organized mesoscale cold front or density current, formed 218 when a cool downdraft spreads out laterally in the form of surface divergence as it hits the surface. 219 Environmental stable-layer wind speeds on the order of 15-20 m s<sup>-1</sup>, and speed profiles exhibiting 220 LLJ structure—a maximum in the lowest several hundred meters—have been known to support 221 the bore formation (e.g., Haghi et al. 2017, 2019). Climatologically over the Great Plains, these 222 conditions are routinely met in the warm season during the frequent occurrences of nocturnal 223 southerly LLJs, for example, Song et al. (2005) found that southerly LLJs occur in 63 % of warm-224 season nights in this region. Thus, when thunderstorms initiate in that region, gust fronts and bores 225 are a regular occurrence, and have been documented in several previous studies (e.g., Koch et al. 226 1991, 2008a,b; Knupp 2006; Loveless et al. 2019, Toms et al. 2017; Mueller et al. 2017; Haghi et 227 al. 2017, 2019; Parsons et al. 2019). Nocturnal Great-Plains thunderstorms are themselves a high-228 frequency occurrence, for example, Geerts et al. (2017) showed that over a six-year summertime 229 period, convective precipitation was observed more than twice per week on average over much of 230 the Great Plains.

231 In surface-tower measurements, bores are most identifiable as an abrupt increase in surface 232 pressure and a windshift toward its direction of propagation. Effects on temperature and humidity near the surface are often minor. Knupp (2006, his Fig.7) shows a pressure bump of ~3 hPa 233 234 accompanied by a drop in wind speed very similar to that depicted here in Fig.1, the wind also 235 veering in time. A major difference between solitary waves (solitons), another common 236 disturbance, and bores is that solitary waves are vertical oscillations where recovery back to pre-237 disturbance conditions is rapid, whereas the effects of a bore persist for some hours. Koch et al. 238 (2008b) observe that in practice, by analyzing field measurement data, "it can be difficult to 239 distinguish bores from density currents and solitons," a notion reinforced by Geerts et al. (2017,

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240 p.778) and Haghi et al. (2017, p.3933). For the present study, making such a distinction is not 241 necessary, but for simplicity and due to the observed slow recovery of the wind speed back to pre-242 disturbance levels, we refer to our ramp-down as a bore in this study. Haghi et al. (2017) found 243 that the frequency of bore activity increased through the night, such that the peak in bore activity 244 was before dawn. Bores tend to move at a faster speed ahead of the gust front that formed them, 245 and often the gust front dissipates, leaving only the bore. Bores can diminish in time as a result of 246 environmental changes, for example if the stability increases above the layer that the bore is 247 propagating through, wave energy can be dissipated upwards, or if the stability of the layer itself 248 changes to become less favorable for bore propagation, either condition can result in the bore 249 weakening.

250 In reflectivity data from weather radar scans, weak lines of enhanced backscatter, referred 251 to as "fine lines," sometimes expand outwards from storm centers (an example will be shown 252 later). Several studies have given examples of radar fine lines, associating them with storm 253 outflow phenomena, including bores (e.g., Knupp 2006; Koch et al. 2008b; Haghi et al. 2017, 254 2019; Mueller et al. 2017; Toms et al. 2017). Thus storm radars can be used to detect fine lines, 255 which is often helpful in short-term forecasting of these kinds of flow disturbances. Many studies 256 have associated bores with nocturnal mesoscale convective systems over the U.S. Great Plains 257 (Blake et al. 2017; Haghi et al. 2019; Knupp 2006; Koch et al. 2008a; Koch et al. 2008b; Parsons 258 et al. 2019). Additionally, bores have been observed over numerous locations worldwide, 259 including Australia (Davies et al., 2017), the UK (Osborne and Lapworth, 2017), Mexico (Martin 260 and Johnson, 2008), and China (Zhang et al. 2020). More details on the formation of internal bores 261 in the atmosphere (Rottman and Simpson 1989) can be found in recent studies (Haghi et al. 2017; 262 Parsons et al. 2019) along with the diagram of flow regimes for two-layer flow over a streamlined 263 obstacle.

Recent research contributions to the understanding of bores and other outflow phenomena have come from the 2002 International H2O Project (IHOP) and the 2015 Plains Elevated Convection At Night (PECAN) project. Both of these studies were primarily aimed at understanding the initiation and propagation of warm-season rainfall and severe-weather events in the Great Plains, important to agriculture and public safety there. Parsons et al. (2017) used IHOP data to study the role of outflow mechanisms in generating new convection, and Stelten and Gallus (2017), and Parker (2021) used case studies from the PECAN dataset for their numerical modeling studies. The generation of new storm activity leads to new storm outflow phenomena,which alter the winds in the rotor layer of wind turbines.

273 From a WE forecasting perspective, it is important to understand that NWP models show 274 poor skill in simulating, and thus predicting, the initiation and movement of moist convection, 275 which includes cumulus clouds that grow into thunderstorms. If the storm initiation is off, then 276 outflow features generated by the storms will be even more poorly modeled, and secondary storm 277 initiation from those outflows, as well as the outflows from those secondary storms, even worse 278 than that. Storm outflows include gust fronts, which significantly increase wind speeds and thus 279 up-ramps in wind-power generation, whereas bores often produce large drops in wind speed, as 280 in the present example, and thus also large drops in wind energy generated. Improving the ability 281 of NWP forecast models to predict storm initiation and development is an important ongoing 282 objective of atmospheric research today.

283 To address these important modeling research goals, these datasets were also used for 284 NWP case studies. In general, the models were able to produce the initiation and movement of 285 storms and the outflows they generated (Stelten and Gallus 2017; Blake et al. 2017; Johnson et al. 286 2018; Parker 2021). Quantitatively, however, the location, timing, and intensity of the storms and 287 outflows have led to significant errors (Gao et al. 2017; Feng et al. 2018). For example, in results 288 from initial, "pristine" storms that were reasonably well simulated, Stelten and Gallus (2017) 289 noted that mean timing errors were 1-1.7 h and location errors, 77-105 km in the first initiation of 290 Plains convection by five models. Commenting on this study, Parker (2021) notes that these 291 models were, "not particularly skillful in terms of the initial timing and location of mesoscale 292 convective system development," a general sentiment expressed in many of the articles 293 mentioned. We note that one forecast model, the Rapid Refresh (RAP), in which the HRRR is 294 nested, addresses this after a storm shows up on the operational radar analysis by adjusting the 295 model's dynamic and thermodynamic fields to account for the existence of the storm (Benjamin 296 et al. 2016). Thus, although the model may miss the initiation of the storms, their effects are 297 represented in forecasts for model runs initialized after the storms appear on operational radars. 298 Overall, the accurate prediction of nocturnal convective systems as well as bores commonly 299 generated by convective outflows (Haghi et al. 2017) in the warm season over the Great Plains 300 remains a challenge for numerical weather prediction models (Zhang et. al, 2019).

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301 In summary, for WE the interest here is in the role of these storms in producing flows that 302 disrupt the normal warm-season wind pattern below 200 m AGL, which on most nights in the 303 Great Plains is dominated by the southerly LLJ. The bulk of the activity for these outflow 304 phenomena is aloft, but they are often associated with large changes in wind speed near the 305 surface, including the wind-turbine rotor layer. Gust fronts can produce significant ramps-up in 306 wind speed, and bores, ramps-down as here. Probably the most important message is that features 307 such as these have their ultimate origins in deep, moist convection (thunderstorms), and predicting 308 convective initiation and subsequent development into rain- or thunderstorms is a difficult forecast 309 for current-generation NWP models.

### 310 3.2 Case Study: Large-scale environment

311 The surface chart for 0900 UTC (Fig. 3) prior to the ramp event in Fig. 1 shows a large-312 scale ridge off the East Coast of the U.S., extending westward to produce southerly wind flow over 313 the southern Great Plains, including the study region in Oklahoma. The 500-hPa chart (Fig.3b) 314 indicates that this southerly flow occupied a deep layer over the region. Such a large-scale 315 subtropical ridge (the "Bermuda high") extending westward from the Atlantic Ocean is typical for 316 August, but Pichugina et al. (2023) have shown that August 2017, the month of LAFE, saw an 317 unusually large number of frontal passages and resulting postfrontal northeasterly wind conditions 318 compared with climatology, including a day-long rain event on 11 August. Here a stationary front 319 stretches west to east at the surface to the north of the study area, separating the southerly flow to 320 the south from cooler air over the northern states. A mesoscale trough passed north to south 321 through the Oklahoma-Texas panhandles, associated with a line of thunderstorms.

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Figure 3. (a) Surface analysis chart for 0900 UTC (https://www.wpc.ncep.noaa.gov/), and (b) 500 hPa chart for 1200 UTC, 21 August 2017. Red arrows show location of SGP study sites.

Composite radar reflectivity images for 0955-1055 UTC (Fig.4) show the most likely source of the major ramp-down in Fig.1. A thunderstorm cell passed north of the study area from west to east, and radar fine lines seen in the images (red arrows) indicate storm gust-front outflow features approaching the study area from the northwest. As described in general in Section 3.1, this fine line is associated with the disturbance seen in Fig.1.





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335	On the study night, the development of the LLJ from 02 to 11 UTC at Site C1 (Fig. 5a)
336	appeared typical of southerly LLJ nights as described by Pichugina et al. (2023), until the ramp
337	event at 1100 UTC; a minor disturbance is also noted starting at ~07 UTC. Doppler-lidar wind
338	speed profiles (Fig. 5c) prior to 1100 UTC show the LLJ maximum speed occurring at 400 m $$
339	AGL, but the depth of the turbulent stable boundary layer (SBL) was less than 200 m, by the
340	Pichugina and Banta (2010) classification, according to which the kink in the profile at that height
341	identifies these as Type 3 profiles (Appendix B). Potential-temperature profiles $\theta(z)$ indicate stable
342	stratification up to 700 m. Profiles after the ramp-down in Fig. 5c, starting with 1100 UTC,
343	indicate large wind-speed reductions up to at least 300 m; then post-sunrise convective mixing is

344 evident in  $\theta(z)$  after 1300 UTC. August sunrises in central Oklahoma are at ~1200 UTC.

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346 Figure 5. Diurnal variability of (a) wind speed and direction from ULID, (b) virtual potential temperature, 347 and water vapor mixing ratio from URLID measurements in the first 1 km AGL on 21 Aug 2017 at C1. (c) 348 5-min profiles of data from both lidars: (black) wind speed, (blue) water vapor mixing ratio (r), (dark 349 yellow) temperature (T), and (dark red) potential temperature ( $\theta$ ) are shown for selected times before (0900-350 1100 UTC), during (1130-1230 UTC), and after (1300-1400 UTC) the observed ramp event. Red asterisks 351 indicate wind speed maxima (LLJ) in the wind speed profiles.

352 Before the ramp, the speed and direction of the LLJ peak developed in a similar manner at 353 the five sites (Fig. 6), but the height of the LLJ maximum was more variable in time and from site to site, as found by Pichugina et al. (2023). The nighttime (0300-1000 UTC) evolution of LLJ 354 355 parameters (Fig. 6a) shows a gradual increase of wind speed maxima (ULLJ) at all sites, slightly 356 stronger at the western (E32, E37) sites by the beginning of the ramp event (just after 1000 UTC 357 at E32; after 1100 UTC at other sites). The heights of the LLJ (Z<sub>LLJ</sub>) were mostly below 400 m 358 except E32 where LLJs were higher after the weaker-disturbance event at 0500 UTC. Wind

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359 direction (ZLLJ) at the windspeed maximum gradually changed from southerly to south-360 southwesterly through the night. Significant wind fluctuations (wind ramp) accompanied by fluctuations in wind direction, temperature, and pressure (Fig. 5a) can be considered as an undular 361 362 bore. An undular bore is a wave or waves in the atmosphere that can be seen on radar or lidar 363 images. These waves can travel across the area of wind farms causing fluctuations in wind power production. Selected profiles within 3 hours before the ramp event (Fig. 6b) illustrate the difference 364 365 between LLJs for all sites and small  $\sim 30^{\circ}$  veering of wind direction with height over the first 1 km AGL. These aspects of LLJ development are typical of a Great Plains summertime southerly LLJ 366 367 (Pichugina et al. 2023).



369 Figure 6. (a) Nighttime (0300-1000 UTC) evolution of LLJ parameters at five sites on 21 Aug 2017: ULLJ 370 - the strongest wind speed in each 10-min wind speed profile below 1 km AGL (or jet nose), ZLLJ - the 371 height of the jet nose, and DLLJ - the wind direction at the jet nose. Colors indicate LLJ parameters for each 372 site according to the legend in the bottom panel. (b) Profiles of wind speed and direction for selected times 373 (0900, 1000, and 1100 UTC) before the ramp-down of wind speed which started around 1030 UTC at E32, 374 1120 UTC at C1<sub>7</sub> and around 1200 UTC at the other 3 sites. Symbols on the top panels indicate LLJ nose 375 (ULL). Profiles of wind speed and direction at E32 were omitted for 1100 UTC since the ramp event started 376 earlier at this site.



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381 through 0800-0900 UTC, just before the major ramp-down event at 1000 UTC. This lesser 382 disturbance, which will be discussed again later, faded as it progressed, as a weaker, shorter-lived 383 lull arrived at the southwesterly Site E37 an hour later, and even weaker disturbances were seen at 384 the other sites after 0700 UTC. At C1 this drop in wind speed was noted at 07 UTC (Fig.5a).



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386 Figure 7. Spatial variability of wind flows between SGP sites: (a) wind speed, (b) wind direction from 387 SLID 10-min measurements on 21 Aug 2017. The larger ramp event was observed at the northern sites E32 and E41 compared to the southern sites E37 and E39. The white areas on each panel indicates missing 388 389 data. The color scale for (a) wind speed is shown up to 20 m s<sup>-1</sup> to reveal the LLJs occurrence before the 390 wind ramp-down event.

### 391 4. Observed wind ramp during LAFE

### 392 4. 1 Time-series of wind flow at C1

393 The August 2017 LAFE study period exhibited a variety of low-level wind flow patterns 394 associated with LLJs, including many nights having northeasterly jets, as described. Lidar 395 measurements (Fig. 8) show significant day-to-day variability of wind speed and wind direction 396 as well as diurnal variability, with stronger nighttime wind magnitudes and larger shear in the first 397 1 km AGL compared to daytime (Pichugina et al. 2023). Modest wind-speed fluctuations that 398 would lead to WE power variation were mostly less than 3 m s<sup>-1</sup>, during the evening transition and 399 at other times on most nights. For example, Fig.8 shows these kinds of routine variations on nights having strong (0-15 m s<sup>-1</sup>) southerly and moderate (4-12 m s<sup>-1</sup>) south-easterly winds (20-25 400 401 August) and on a weaker-wind night (30 August). The variations of wind speed during these days 402 and over the study period were relatively small in magnitude and much shorter in time compared

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to the major ramp event observed on 21 August. Thus, these fluctuations would unlikely present
a serious risk to wind-farm operations, because wind power (P) was still being generated (P>0),
although at a reduced, hard-to-predict level.
On 21 August much larger wind-speed fluctuations were observed at C1 at all heights
across a typical wind-turbine "rotor layer" of 53-138 m and higher, up to 200 m (Fig. 8, 21 Aug).
Several periods of small ramping (~2-3 m s<sup>-1</sup>) and relatively constant shear are seen before the

409 larger ramp-down (> 15 m s<sup>-1</sup>) event that started around 11 UTC and lasted about 5 hours, 410 including the daytime recovery back to 10 m s<sup>-1</sup> southerly flow. Wind directions before this event 411 were mostly from the south, then veered ~120 degrees through a deep layer during the event, as 412 in the bore studied by Knupp (2006).



Figure 8. Time series of (a) wind speed and (b) wind direction from ULID measurements at C1 are shown for selected days (20, 25, and 30 Aug), and for 21 Aug 2017 when a large ramp event was observed. Winds are shown at several heights through the 53-202 m layer and indicated by colors according to the color table. Range of wind speed and prevailing wind directions at each panel is provided for 0-15 UTC.

420 Wind-speed and wind-direction ramps at C1 on this day were also observed by in-situ 421 measurements averaged over 30-min intervals. At 1100 UTC sonic anemometer measurements at 422 25 and 60 m AGL on the 80-m meteorological tower (Fig. 9b) indicate the ~5-hour ramp-down-423 and-recovery event with 5 m s<sup>-1</sup> and 7 m s<sup>-1</sup> decreases in wind speed. Even lower, cup-anemometers 424 at 2.5 and 10 m AGL from three flux stations located near but not at C1 (Fig. 9a) indicate two small ramp-down events of wind speed (~ 3 m s<sup>-1</sup>) and wind direction at ~0700 and 1300 UTC. 425 426 After 1000 UTC a rise in surface pressure of more than 2 hPa (Fig.9c) marked the beginning of 427 the major drop in wind speed of the bore-generated, ramp-down episode, an effect also noted in

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other bore case studies (e.g., Knupp 2006; Koch et al. 2008; Toms et al. 2017; Blake et al. 2017).

Sudden veering of the wind (to westerly or northwesterly) and small fluctuations (flux stations) or

drops (80-m tower) in temperature are also consistent with the previous bore examples. The

smaller event evident at E32 at 05 UTC can be seen as a drop in speed in the 80-m tower

measurements here at ~07 UTC, and a drop in speed and a wind shift at the flux-tower sites.

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Figure 9. Time-series of a wind speed, direction, and temperature on 21 Aug from (a) cup anemometer
measurements from energy balance flux stations (T1-T3) at (open circles) 2.5 m and (filled circles) 10 m;
(b) 30-min sonic anemometer measurements at 25 m and 60 m on an 80-m meteorological tower. (c)
Barometric pressure at 10 m at the flux stations in (a).

### 442 4.2. Vertical structure of the wind flow

443 Time-height cross sections and profiles of wind speed, direction, temperature, and 444 humidity have been presented in Fig. 5. The combined effects of data availability from ULID and 445 URLID at C1 allow us to investigate the diurnal variability of key variables in the first 1 km AGL 446 and examine the changes in profiles of these variables before, during, and after the ramp event 447 (Fig. 5c). As noted, development of the LLJ prior to the event was typical of a southerly-LLJ night, 448 a deep LLJ forming a peak of 20 m s<sup>-1</sup> at 400 m and a stable  $\theta$  profile up to 700 m.

449 During the event (Fig. 5c, 1100-1200 UTC profiles), the drop in wind speed at C1 was 450 especially dramatic between 200 and 300 m AGL, and the wind shifted to a west-northwesterly 451 direction, more directly aligning with the direction of propagation of the bore. The spatial 452 variability of the ramp-down among the four extended sites (Fig. 7) appears as a more significant 453 reduction of wind speed and a larger wind-direction veer at northern sites (E32 and E41) compared 454 to the southern sites (E37 and E39). Differences in the magnitude and timing of the wind-speed ramp event are evident among sites (Fig. 7a), but some similarities can be noted such as the 455 456 occurrence of the LLJs before the ramp-down in the lowest several hundred meters. The wind 457 direction (Fig. 7b) below 400 m at all sites, that had been south-southeasterly  $(150^{\circ})$  before the 458 ramp, changed to south-westerly and westerly during the ramp for a short period at the northern 459 sites. The departure of the bore and the restoration of stronger southerly flow after 1300 UTC 460 occurred in conjunction with the onset of daytime heating and vertical mixing after sunrise, the 461 effects of which can be seen in the wind and  $\theta$  profiles in Fig. 5c.

462 The relative humidity and potential temperature retrieved from the Atmospheric Emitted 463 Radiance Interferometer (AERI) at the five sites (Fig. 10a-b) indicate a moistening and a 5-6 C<sup>o</sup> 464 cooling across a layer near the surface several hundred meters deep during the ramp event. The 465 moistening and cooling are more evident over E32 and C1 compared to the others sites, and coincide with a stronger reduction in the near-surface potential temperature gradient structure as 466 467 shown in Figure 10c. These changes reflect the proximity of these sites to the storm and the 468 propagation direction of the outflow-generated bore northwest of the ARM site (recall Figure 4), 469 and illustrate the impact of a passing bore on the stability structure as well as the disruption of a

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470	strongly sheared flow (Figures 5 and 7). The relatively larger decrease in stability at E32 and C1 $$
471	that coincides with deeper moist layers extending a kilometer or so deep in Figure 10a suggests
472	stronger turbulence production associated with the passage of the bore. The rapid vertical
473	displacement and overturning that destabilizes the low-level atmospheric structure resembles to
474	the results of Blake et. al. (2017), particularly the moistening near the surface. Furthermore, the
475	larger reduction in the potential temperature gradient within the 100 m to 400 m layer indicates a
476	stronger response near the surface, which would affect more of the rotor layer.



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### 483 5. Spatial variability of winds during observed ramp event

484 Comparing measurements from lidars at the five SGP sites (Fig. 2) reveals wind-flow 485 variations due to distances between instruments resulting in measuring atmospheric variables under different wind flow regimes and surface properties. During LAFE (Pichugina et al. 2023) 486 487 differences in LLJ wind-speed measurements in general were evident between west (E32, E37) 488 and east (E39, E39) sites as well as between south (E37, E39) and north (E32, E41) sites depending 489 on the wind direction. Here the major differences are due to the propagation of the bore through 490 the measurement array.

### 491 5. 1 Time series at 5 sites

492 Time-series (Fig. 11) of wind speed and wind direction from lidar measurements at several 493 selected heights from ~100 to 700 m AGL illustrate differences among the five sites as well as 494 shear between heights at each site.





498 measurements at C1 and SLIDs measurements at the 4 extended facilities (E32-E41) according to the 499 legends. ULID data (2Hz) is averaged over 10 min to fit the time resolution of SLIDs. Panels for each site are presented in Fig. 11 according to the site location as shown on the embedded USGS map at the top of 501 the figure.

502 At C1 the largest reduction of wind speed (11.3 m s<sup>-1</sup>) within the ramp-down period (1000-1300 UTC) was observed at 456 m AGL. The wind direction at the two lowest heights (95 and 503 504 117 m) significantly veered during 1100-1200 UTC changing from southerly (180º and 184º) to 505 north-westerly (317º and 312º); then, within the next 30 min, it shifted to 128º and 178º (Tab. 2). 506 The strongest wind shear was observed in the LLJ period before the ramp event, reaching 0.061 s<sup>-</sup> <sup>1</sup> between 96 and 117 m at 0500 UTC and  $0.05 \text{ s}^{-1}$  between 222 m and 95 m at 0545 UTC. Overall, 507 508 the shear observed between all heights before and during the event was strong, comparable to 509 previous studies that found values up to 0.1 s<sup>-1</sup> observed from lidar measurements in Kansas (Banta 510 et al. 2003) and southeastern Colorado.

511 Significant wind ramps at the northern sites (E32 and E41) and C1 compared to the 512 southern sites (E37 and E39) is demonstrated across different heights in Figure 12. Following the 513 approximate propagation direction of the bore, the ramp event is first observed at E32, an hour 514 later at C1, and 30 minutes later at E41 relative to C1. The stronger ramp event at E32 was 515 preceded by large pre-ramp wind shear and an additional smaller wind ramp around 05 UTC. The largest down-ramp of wind speed (14.7 m s<sup>-1</sup>) was observed at 456 m at E32, which was closest to 516 517 the storms that produced outflows that led to the bore (recall Figure 4). The wind speed at E32 dropped from 20.7 m s<sup>-1</sup> at 1000 UTC to 5.9 m s<sup>-1</sup> at 1145 UTC. The wind direction at several 518 519 heights changed from southerly (184-198°) at 1000 UTC to north-westerly (308-284°) at 1030 UTC, then backing to south-westerly (207-227°) at 1100 UTC. The bore diminished in amplitude 520 521 as it propagated to the southeast, along with smaller reductions in wind speed and the wind-522 direction veer. The decreased impact from the bore to the southeast agrees with the discussion on 523 moist-stability characteristics shown in Figure 10.

At the southern sites (E37, E39) wind speeds declined more gradually from 1100 to 1500 UTC with a slight increase in directional shear during this period. At E37 the short (~20 min), smaller down ramp previously discussed was observed around 0530 UTC at the two lowest levels with a 5.3 m s<sup>-1</sup> change in wind speed and wind direction veer from 170° to 253°.

528A clearer view of the propagation of the ramp through the measurement array can be529obtained by plotting the time series at a given height for all sites, as shown in Fig. 12.

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Figure 12. Time series of (a) wind speed and (b) wind direction at several heights from lidar measurements at the five sites are shown by colors according to the legend at the top right panels. Arrows at the panels for 223 m illustrate the beginning of the wind speed drop and veer of the wind direction at this particular height.

The largest response in wind speed appears in the 300-500-m AGL layer, diminishing with height until little evidence of a disturbance can be seen at 817 and 922 m AGL. The magnitude and timing of the wind ramp varied from height to height and between sites. The timing difference between sites at the beginning of the wind ramp is illustrated by the arrows on the panels for 223 m. The initial drop in wind speed and shift to west-northwesterly flow is seen at 1000 UTC at the northwest site E32, then in diminished form at the southwest site E37. The significant drop in speed and shift to westerly noted previously at C1 occurred at 1100 UTC, followed by the northeasterly site E41. Finally, the ramp passed through the southeastern site E39 starting at 1330
UTC, but as with E37, little veering of the wind occurred.

543 In some studies (Ahn and Hurl, 2022; Gallego et al. 2014) ramp events have been 544 characterized by the following parameters: ramp start and end times, duration, ramp speed 545 minimum, and the magnitude of the change in wind speed, which is the difference between the 546 maximum and minimum of wind speed ( $\Delta$ sp) or wind direction veer ( $\Delta$ dir) during the ramp-down 547 event. Two of these parameters ( $\Delta$ sp and  $\Delta$ dir) are shown in Fig. 13 for several heights at each site. The largest magnitude of wind-speed ramp-down (Fig. 13a) was observed at E32 (14.7 m s<sup>-1</sup> 548 549 at 455 m) followed by the drop at E41 (12.6 m s<sup>-1</sup> at 350 m) and at C1 (11.3 m s<sup>-1</sup> at 455 m). The largest wind direction veer was observed at C1 (200°) and E32 (160°) at 90 m (Fig. 13b). 550



551Height, m AGL552Fig. 13. Magnitude of (a) wind speed ramp-down and (b) wind direction veer at 8 heights through the553layer of 90-714 m from 10-min lidar measurements at the five ARM SGP sites.

### 554 5.2 Estimate of wind-power loss due to the observed wind ramp

555 Strong increases or decreases in wind speed over a few hours lead to a corresponding 556 ramping in wind power production. The impact of wind ramps on wind power-plant output has 557 been studied in recent decades (Dalton et al. 2012, Galego et al. 2014; Lee et al. 2012; Pichault et 558 al. 2021; Smith and Ancell, 2017; Wharton et al. 2008; Yang et al. 2013; Zhang et al. 2014). A 559 high correlation was found between the power computed from measurements by three scanning lidars located in complex terrain, separated from each other by 30-40 km, and the total power 560 561 generated over the BPA area during episodes of marine intrusions when the winds were 562 consistently westerly (Pichugina et al. 2020). The influence that wind ramps can have on power output can be estimated for a "hypothetical" wind turbine with ~90 m hub height and 70-m rotor 563

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diameter (Fig. 14). Wind power calculated as  $P = \frac{1}{2} * \rho \pi R^2 S^3 C p_j$ , where  $\rho$  is the air density, R=35564 565 m is the blade length or the radius of the area swept by the turbine, S is measured hub-height wind speed, and  $Cp_i$  are coefficients set to represent the power curve with 4 m s<sup>-1</sup>, 25 m s<sup>-1</sup>, and 12 m s<sup>-1</sup> 566 567 <sup>1</sup> cut-in, cut-out, and rated wind speed respectively (GE Energy, 2009). The cut-in and cut-out thresholds may vary between power curves of different turbines but in common represent wind 568 569 speed at which the turbine starts generating electricity or wind speed at which turbine can start spinning so fast and can be damaged. Wind speeds ranging between these two extreme points 570 571 (rated winds) are favorable for turbine operations.

Fig. 14 shows that during the ramp event on 21 August, no power will be generated at the northern sites (E32, E41) and the central facility (C1) for 1h 30 min. At the southern sites (E37, E39) winds do not decrease below 4 m s<sup>-1</sup> during this period and some power will still be generated. The smaller ramp event was also observed at E37 around 0530 UTC or local midnight (0030 CDT - Central Daylight Time), and the decrease of wind speed below the cut-in threshold would have led to zero power for about 30 min (0530-0600 UTC) for turbines near that site.



Figure 14. Time-series of wind speed from lidar measurement and computed power for a "hypothetical"
wind turbine with 70 m rotor diameter are shown for (a) central (C1) and northern (E32, E41) sites, and (b)
southern (E37, E39) sites. Wind speed is shown for the lowest height (91 m) of lidar measurements at E32E41 and the closest height (95 m) from ULID at C1. (Time UTC=CDT+5h). Brown arrows point to periods
of zero power.

584 Similar to the wind speed, a positive change in generated electrical power over short time 585 intervals is defined as a "ramp-up", whereas a negative change is referred to as a "ramp down" 586 (Ahn and Hurl, 2022; Gallego et al. 2014). The ramp parameters for wind speed and the computed 587 power are shown in Table 2 for all five sites. The last column shows these parameters for the 588 smaller ramp event observed at E37 around 0500 UTC.

589 590

Table 2. Ramp parameters from lidar-measured 90-m wind speed at 5 SGP sites and the period when computed wind Power (MW) equals 0.

Ramp	C1	E32	E37	E39	E41	E37
parameters						
		Ov	erall event time	e (UTC)		
Start	1030	1030	1240	1300	1100	0500
End	1600	1500	1400	1500	1600	0630
Duration	0530	0430	0240	0200	0500	0130
		Ra	mp down time	(UTC)		
Start	1030	1030	1240	1300	1100	0500
End	1240	1130	1310	1400	1240	0545
Duration	0210	0100	0030	0100	0140	0045
		F	amp up time (	UTC)		
Start	1240	1130	1310	1400	1240	0545
End	1600	1500	1400	1500	1600	0630
Duration	0320	0330	0110	0100	0320	0045
Power=0						
Start	1130	1130	1300	1400	1140	0530
End	1500	1330	1350	1420	1430	0550
Duration	0330	0200	0050	0020	0250	0020

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### 595 6. HRRR evaluation: Forecast of the wind ramp

596 Forecasting of ramp events that occur over short temporal and spatial scales can be a 597 difficult task, and accurate simulations of the synoptic processes leading to these dynamic changes 598 of wind speed require models having sufficient temporal and spatial resolution (Koch et al. 599 2008a,b; Yang et al. 2013). In this paper, prediction of the observed wind ramp event of 21 August 600 by NWP models was also explored. Simulations of wind flows over the ARM SGP area on that day were available from the operational HRRRv2 and the experimental HRRRv3 (which become 601 602 operational in July 2018). An essential feature of the HRRR system is its rapid (hourly) updating, 603 useful for assimilating the latest weather data. With 3-km grid spacing, the hourly-updated HRRR 604 model provides the opportunity to better represent convection and its associated hazards. It is 605 widely used for severe-weather, renewable-energy generation, and flash-flood forecasting (Dowell 606 et al., 2022; James et al. 2022). Given the many difficulties in simulating storm initiation, HRRR's 607 hourly update cycle makes it especially advantageous in overcoming many of the problems NWP 608 models have in general in predicting nocturnal convection and the accompanying/ mesoscale 609 phenomena. Both model versions provide forecasts with a relatively fine temporal (hourly) and 610 spatial (3-km grid horizontal grid) resolution. Detailed descriptions of physics and 611 parameterization schemes for all HRRR versions can be found in Dowell et al. (2022) and James 612 et al. (2022).

613 In this study, HRRRv3 was evaluated against Doppler lidar measurements at five SGP sites 614 to address how well the model simulates the spatial and vertical variability of the wind profiles, to 615 quantify the ability of the hourly HRRRv3 outputs to capture wind ramps, to estimate model skill 616 for several forecast lead times, and to evaluate the forecast performance for different BL conditions 617 such as "ramp-day" versus "no-ramp day".

Fig. 15 gives a time-height cross-section overview of ULID-measured and modeled windspeed forecasts for the 06-18 UTC time interval surrounding the ramp event, showing how model skill changed with increasing forecast lead time. The selection of initial times for both days is based on the availability of the archived HRRRv3 model outputs (00z on 21 Aug is missing) and the maximum day-ahead forecast lead time (Table 1). For comparison, the last column (Fig. 15c) shows wind speed forecasts on 21 Aug from the operational HRRRv2 model.

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Figure 15. Wind speed from (topmost panels) hourly averaged ULID measurements at C1 and (lower panels) model forecasts valid for 0600-1800 UTC on 21 Aug. (a) HRRRv3 forecasts on 21 Aug for initial times 01z (fcsts 5-18), 03z (fcsts 3-16), 05z (fcsts 1-13), and 06z (fcsts 0-12). (b) HRRRv3 day-ahead forecasts on 20 Aug for initial times 00z (fcsts 30-42), 06z (fcsts 24-36), 09z (fcsts 21-33), and 12z (fcsts 18-30) valid for 21 Aug. (c) Same as (a) but from the operational HRRRv2 model.

All runs show a well-developed LLJ by 0600 UTC. The 00-12-h lead-time forecasts initialized at
0600 UTC (Fig. 15a) do not indicate a ramp-down disturbance, but forecasts initialized 1-, 3-, and
5-h earlier did show drops in wind speed below 300 m within the time interval displayed, although
the timing was earlier than observed. Significant reductions of wind speed on 21 Aug are seen in
HRRRv3 forecasts initialized at 01z and 05z, but beginning 2-3 hours earlier than measured.
Forecasts initialized at 03z show reduced winds (1-3 m s<sup>-1</sup>) above 400 m which are not indicated

637 by measurements. Comparing HRRRv3 forecasts initialized at 03z on 20 Aug (day-ahead, Fig. 638 15b) and 21 Aug (same-day), the day-ahead runs (Fig. 15b) initialized from 20 Aug valid for 0600-639 1800 UTC on 21 Aug show strong winds above ~200 m with developed LLJ up to 1200 UTC but no disturbances below 200 m. Overall, no indication of the ramp event was found from all lead 640 641 times (every 3 hours from 00z) for day-ahead forecasts. The operational HRRRv2 (Fig. 15c) 642 shows some wind drop for the 05z- and 06z - times runs at 0800 UTC and 0900 UTC respectively, 643 but shows no wind-ramp indication from other initial times such as 01z and 03z.

644 As pointed out in Bossavy et al. (2010), forecasts using large-temporal-scale information 645 about ramps may lead to a significant time delay "resulting in turn to the so-called phase error". 646 But they proposed that using ensembles to generate confidence intervals may produce better 647 forecasts of ramp timing with more reliable confidence intervals for each look ahead time.

648 Time series (0600-1800 UTC) of wind speed at three lidar-measurement heights and 649 HRRRv3 forecasts for three initial times (Fig. 16a) on 21 August demonstrate better agreement at 650 southern sites (E37, E39), compared to the northern sites (E32, E41) and the central facility (C1) 651 where the ramp event was most evident. Wind-speed reduction at northern sites was captured by 652 01z and 05z forecasts but the significant time and vertical offsets led to large errors. It is clear that 653 significant fluctuations occurred at the northern and central sites, and the NOAA experimental 654 HRRR (HRRRx) simulated radar reflectivity maps (Fig. 20b, Appendix A) show that a 655 thunderstorm cell is represented in the simulation, and is the likely source of these wind 656 disturbances. Significant timing errors in the wind ramps are noted.

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Figure 16. (a) Time-series of wind speeds on 21 Aug are shown for Northern sites (E32, E41), Central facility (C1), and Southern sites (E37, E39). Lidar data (black lines) are shown for 3 heights that are slightly different for ULID at C1 and SLIDs at E32-E41 as indicated on each panel. HRRRv3 outputs from (dark red) 01z, (red) 05z, and (blue) 06z are linearly interpolated to lidar heights at each site and shown for the forecasts valid for 0600-1800 UTC. (b). Same as Fig. 16a but for 20 Aug. Red lines show forecasts for initial time 04z because 05z is missing.

For comparison, time series of wind speed at 3 heights are shown for 20 Aug (Fig. 16b), to illustrate a better agreement between lidar measurements and HRRRv3 simulations at all five sites on a day without any significant wind fluctuations. Figure 16 shows that time offsets of the wind down-ramps between lidar and model produce larger errors at C1, followed by errors at the northern sites E32 and E41. Large errors at E37 at 0600 UTC can be explained by the offset in the forecast of the short ramp observed by lidar at this site during 0520-0610 UTC (Table 3, last column).



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673 Figure 17. Time-series of wind speed (a) difference ( $\Delta$ , m s<sup>-1</sup>) between lidar measurements and HRRRv3 674 05z forecast at 5 sites on 21 Aug 2017. (b) Absolute difference ( $\Delta$ , m s<sup>-1</sup>), and (c)  $\Delta$  speed normalized by 675 the wind speed from lidar. Time series are shown for the 3 closest heights of ULID at C1 and SLIDs at 676 E32-E41 as indicated by the legend.

677 Mean absolute wind speed difference (abs  $\Delta$ -speed) at all three heights (Fig. 18a), averaged 678 over the period of the observed ramp event at all five sites (09-13 UTC), is also larger at the 679 northern sites (E32, E41) and C1 compared to the southern sites (E37, E39), whereas the average 680 over a longer period (0600-1800 UTC) shows little difference between northern (E32, E41) and 681 southern (E37, E39) sites (Fig. 18b). In comparison, mean  $\Delta$ -speed for these periods on 20 Aug 682 are significantly smaller (Fig. 18c, d).

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Mean ABS ∆speed, 21 Aug

Figure 18. Absolute difference ( $\Delta$  speed) between HRRRv3 and lidar measurements at five sites averaged over (a) 09-13 UTC and (b) 06-18 UTC on 21 Aug. Forecasts are from the initial time 05z. Same on the bottom panels (c) and (d) but on 20 Aug for 04z forecasts.

### 688 8. Conclusions

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689 Changes in the supply of power from wind-generation facilities are a significant issue for 690 the wind energy community. Unpredicted down-ramps are especially problematic, because 691 expected power is suddenly unavailable to the power grid. A significant ramp-down of wind speed, 692 which would have resulted in an abrupt loss of wind power for more than two hours, was observed 693 over north-central Oklahoma during the August 2017 LAFE project. We attributed the ramp to a 694 bore, most likely produced by a gust front from thunderstorm activity to the northwest. It passed 695 over the five-site SGP network of Doppler lidar wind-profiling sensors, allowing the spatial and 696 temporal characteristics of the disturbance to be studied. The high temporal and vertical resolution 697 of Doppler lidar wind profiles made it possible to reliably determine a ramp event observed on 21 698 Aug 2017 at the ARM SGP sites in central Oklahoma, USA, and analyze the site-to-site variability 699 of the ramp parameters. A significant down-ramp of wind speed passed through the northern and

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central sites but stalled before reaching the southern sites. The bore caused a significant drop ofwind speeds at these sites to below cut-in values of wind turbines in this area.

702 The ability to accurately predict ramps-down such as found in this study is an important 703 forecasting challenge for wind energy. Predicting the onset of deep moist convection is a difficult 704 problem for current generation NWP forecast models. Even when thunderstorm cells are 705 represented in about the right place in models, the resulting outflows and preceding wave activity, such as the bore in this study, are subject to large errors in timing and other properties of the wind 706 707 structures, indicating the need for further research into these systems. Supplementary, nested 708 arrays of wind sensors to detect these flows in real time and extrapolate their movement (Banta et 709 al. 2013), or use of wind-turbine mounted anemometers upstream of the disturbances, are other 710 potentially important resources for predicting these ramps an hour to a few hours ahead of time.

### 711 Acknowledgments

712 This study was conducted as part of the Land-Atmosphere Feedback Experiment (LAFE). 713 This research was supported in part by NOAA cooperative agreements NA17OAR4320101 and 714 375 NA22OAR4320151, the U.S. the U.S. Department of Energy's Atmospheric System 715 Research, an Office of Science Biological and Environmental Research program, under Grant No. 716 DE-SC0020114, as part of the Atmospheric Radiation Measurement Program and Atmospheric 717 System Research Program, and the NOAA Oceanic and Atmospheric Research Office of Weather 718 and Air Quality, the NASA Water and Energy Cycle Program, the German Federal Ministry of 719 Education and Research (BMBF), and the University of Hohenheim. We gratefully acknowledge 720 the efforts of the ARM Southern Great Plains managers and PNLL staff for maintaining and 721 operating that site both during this campaign and over the last three decades. Lastly, the scientific 722 results and conclusions, as well as any views or opinions expressed herein, are those of the authors 723 and do not necessarily reflect the views of NOAA or the Department of Commerce.

### 724 Data Availability Statement

Datasets from scanning Doppler lidars at 5 SGP sites used during this study are openly available from the Atmospheric Radiation Measurement (ARM) Southern Great Plains (SGP) Archive at https://www.arm.gov/capabilities/observatories/sgp\_and in the user manual\_compiled by T. Shippert, R. Newsom, and L. Riihimaki. ARM Data Center: Atmospheric Radiation



- Measurement (ARM) user facility. 2016. Doppler Lidar Horizontal Wind Profiles
  (DLPROFWIND4NEWS). 2021-06-01 to 2021-08-31, Southern Great Plains (SGP) Waukomis,
- 731 OK (Extended) (E37). Data set can be accessed at <u>http://dx.doi.org/10.5439/1178582</u>.
- 732 The Land Atmosphere Feedback Experiment. Datasets from the Land Atmosphere Feedback
- 733 Experiment (LAFE, 2017) are available via the ARM Data Discovery portal at:
- 734 https://adc.arm.gov/discovery/

### 736 APPENDICES

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### 737 Appendix A. Synoptic analysis of BL conditions on 21 Aug

738 We have characterized the rapid changes of all variables during ~11-14 UTC on 21
739 August as a bore, generated ahead of a gust front from thunderstorm activity to the northwest of
740 the measurement sites. Here we show supplementary analyses of some meteorological conditions
741 that from the context of this scenario, and some HRRR output for the period of interest.

### 742 A.1 Temperature and humidity from North American Regional Reanalysis (NARR)

743 The North American Regional Reanalysis (NARR) is a model, produced by the National 744 Centers for Environmental Prediction (NCEP), that generates reanalyzed data for temperature, 745 wind, moisture, soil, and dozens of other parameters. The NARR model assimilates a large amount 746 of observational data from a variety of sources to produce a long-term picture of weather over 747 North America. NARR 3-hourly composites of Vector Wind and Precipitable water (Fig. 19a, 748 Appendix A) indicate the southerly flow over the region and a strong peak in water vapor to the 749 north of SGP. Stronger (15-18 m/s) winds over the 09-12 UTC composite (middle panel) diminished by 5-6 m s<sup>-1</sup> for the 12-15 UTC composites while wind direction changed from 750 751 southerly to south-westerly. Maps of the Precipitable water (Fig. 19b, Appendix A) show drier 752 conditions in the vicinity during 09-12 UTC. Overall, the ramping event on Aug. 21 occurred 753 during a relatively dry period with the most significant recent rain observed in this area 10 days 754 earlier (Pichugina et al. 2022).

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### 759 A.2 Wind speed from the Experimental High Resolution Rapid Refresh (HRRR<sub>X</sub>) model

760 HRRRx is the experimental and advanced version of HRRR developed and constantly updated by 761 NOAA/GSL. The version used in this paper is from HRRRv3 (2016-2017), which became 762 operational at the National Weather Service (NWS) on 12 July 2018 with CONUS, Alaska domain 763 coverage. The major changes for HRRRx included improvements in the MYNN PBL scheme 764 (addition of a mass-flux scheme, transition to EDMF framework); and a hybrid vertical coordinate. 765 The major changes for data assimilation: improvements to better retain stratiform clouds; reduced 766 latent heating for radar-identified moist-convective cells introduced into RAP (Dowell et al. 2022; 767 James et al. 2022). HRRRx simulation of 15-min Winds at 80 m and Reflectivity taken for the initial time 3z (Fig. 20a, Appendix A) illustrate the reduction of winds at 1200 UTC compared to 768 769 winds at 0900 UTC, and the main convective cell moving fairly quickly from the SW to the NE, 770 with substantial activity within an hour or two of the observed ramp timing. The analysis of the 771 NOAA next-generation radar (NEXRAD) images and satellite surface maps (Fig. 20b, Appendix

Fig. 19, Appendix A. NARR 3-h composites (averages) of the hourly mean (a) vector wind and (b)
 precipitable water are shown for periods before (06z), during (09z), and after (12z) the event. The figure
 was created using *visualization tools at* <u>https://psl.noaa.gov/data/narr/</u>

A) confirms that the convention propagated to the north of the study area. The observed ramp
event, larger at the northern sites can be attributed to this convention but the detailed
characterization of its propagation is out of the scope of this paper.





Figure 20, Appendix A. (a) 80-m winds from NOAA/GSL's experimental 3-km HRRRx simulation with
15-min temporal resolution are shown for 0900 UTC (before ramp event) and for 1200 UTC (during ramp
event). SGP lidar sites are indicated by the dark red circles. (b) HRRRx 1-km AGL Reflectivity for the
same time as in (a).

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### Appendix B. The low level jet (LLJ) and basic types of mean wind speed profile observed during previous studies in Great Plains.

The atmospheric flow phenomenon known as the LLJ is a maximum in the boundary layer wind profile (Fig. 21a), frequently observed during warm months throughout the Great Plains of the United States (Bonner 1968; Mitchell et al. 1995; Whiteman et al. 1997; Banta et al. 2003; Song et al. 2005;) Typically LLJ begin to develop around sunset in fair weather conditions, reach peak intensity a few hours after midnight, and dissipate with the onset of daytime convective mixing.

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Fig. 21, Appendix B. (a) Definition of the LLJ parameters where the red point indicates the maximum of wind speed ( $U_{LLJ}$ ) and the height of this maximum ( $Z_{LLJ}$ ).  $D_{LLJ}$  is the corresponding wind direction. (b) Categories of wind profiles that were frequently observed from lidar measurements during the previous experiments in the Great Plains (from Pichugina et al, 2010; ©AMS). Type 1 wind profiles are the classic LLJ shape with a distinct maximum or "nose". Type 2 wind profiles represent a uniform or "flat" profile above the shear layer. Type 3 represents wind profiles in which the shear in the subjet layer (and usually the variance profile as well) showed a layered structure.

Data from Doppler sodars, lidars, and other high-resolution observational platforms indicate that peak LLJ winds are often found within 500 m of the ground (Whiteman et al. 1997; Banta et al. 2003; Song et al. 2005). The classic LLJ wind profiles exhibits a distinct maximum or "nose," with wind speed (U<sub>LLJ</sub>) decreasing both above and below a distinct maximum (Fig. 21a). Increased LLJ winds at night can be an important resource for wind turbine operations in the U.S. Great Plains.

804 Several mechanisms have been proposed to explain LLJ accelerations in the Great Plains 805 (Blackadar, 1957; Holton, 1967). Here LLJ is taken to mean the vertical layer of the previous

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806	afternoon's unstable boundary layer that accelerates in response to the nighttime surface cooling
807	(Blackadar,1957). Profiles in this accelerated layer can assume different shapes (Banta et al. 2002,
808	2006), including the classic LLJ profile, a uniform or "flat" profile, and a layered structure profile
809	as depicted in Fig. 21b. This nocturnal wind acceleration produces a layer of strong shear adjacent
810	to the earth's surface, which generates turbulence. The significant shear-generated turbulence
811	within the layer of wind turbine blades may influence turbine operations.

### 812 Appendix C. Composite reflectivity from NEXRAD at 0455 and 0555 UTC.

- 813 The smaller disturbance (Fig. 22, Appendix C) was observed just before the major ramp-down
- 814 event studied in this paper. This lesser disturbance created a significant but shorter-lived ramp
- 815 event noticed at all sites with the more noticeable at the southwesterly Site E37 around 0500 -
- 816 0630 UTC (Fig.14b).



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- Fig. 22, Appendix C. Composite reflectivity from NEXRAD (Next Generation Radar) 1km
  MOSAIC on 21 Aug 2017 at (a) 0455 and (b) 0555 UTC. White dots denote approximate
- 820 locations of SGP measurement sites
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### 822 Appendix D. Mean wind speed difference between HRRRv3 and lidar measurements

Table 3, Appendix D. Mean over 06-15 UTC wind speed abs difference between HRRRv3,

824 initial times (1z-6z) and lidar data at several heights. Results are shown for smaller ramp days

825 (20, 25, and 30 Aug) in comparison to the significant ramp event observed on 21 Aug.

Date	Lead	Height, AGL					
	time	91 m	117 m	137 m	227 m	350 m	45627
	1z	0.91	0.83	0.80	1.29	1.43	1.62
	2z	1.31	1.27	1.27	1.59	1.32	$1.42^{828}$
20 4 110	3z	1.32	1.19	1.12	1.36	1.18	1.279
20-Aug	4z	1.19	1.10	1.07	1.40	1.17	1.27
	5z	-	-	-	-	-	830
	6z	1.70	1.64	1.63	1.92	1.32	1.44
	1z	-	-	-	-	-	
	2z	0.60	0.63	0.78	1.08	1.37	1.892
25 1 110	3z	0.92	1.08	1.22	1.54	1.76	2.18
25-Aug	4z	0.91	1.06	1.21	1.39	1.46	$1.74^{-833}$
	5z	0.94	1.10	1.29	1.62	1.73	2.064
	6z	0.70	0.86	0.93	1.50	1.63	1.70
	1z	1.55	1.51	1.52	1.08	1.09	1.805
	2z	-	-	-	-	-	ōac
20 1 110	3z	1.46	1.30	1.26	0.70	0.72	0.98
30-Aug	4z	1.72	1.56	1.41	0.84	0.58	0. <b>§</b> 37
	5z	1.40	1.21	1.09	0.56	0.38	0.63
	6z	1.00	0.84	0.73	0.38	0.59	0.968
	1z	2.79	2.79	2.76	2.43	1.93	1.880
	2z	2.69	2.37	2.09	2.10	5.40	7.34
21-4119	3z	2.20	2.50	2.79	3.94	4.60	5. <b>84</b> 0
21-Aug	4z	2.26	2.47	2.65	3.46	4.86	5.73
	5z	3.08	3.17	3.22	2.97	2.78	2.39 <sup>1</sup>
	6z	3.19	3.43	3.68	4.03	3.46	2.842

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HRRRv3 outputs for some initial times were missing in the NOAA/GSL archive but the
corresponding results are clearly show much smaller errors for days with smaller wind ramps.
Errors at 91 m, approximately the hub-height of the most turbines in this area, range between 0.60
and 1.72 m s<sup>-1</sup> on these 3 days in comparison to 2.20-3.19 m s<sup>-1</sup> on 21 Aug.

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