



1 **Evaluating WRF-GC v2.0 predictions of boundary layer and vertical ozone**
2 **profiles during the 2021 TRACER-AQ campaign in Houston, Texas**

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14 **Abstract.** The Tracking Aerosol Convection Experiment Air Quality (TRACER-AQ) campaign
15 probed Houston air quality with a comprehensive suite of ground-based and airborne remote
16 sensing measurements during the intensive operating period in September 2021. Two post-frontal
17 high-ozone episodes (September 6-11 and 23-26) were recorded during the said period. In this
18 study, we evaluated the simulation of the planetary boundary layer (PBL) height and the vertical
19 ozone profile by a high-resolution (1.33 km) 3-D photochemical model, Weather Research and
20 Forecasting (WRF)-driven GEOS-Chem (WRF-GC). We contrasted the model performance
21 between ozone-episode days and non-episode days. The model captures the diurnal variations of
22 the PBL during ozone episodes ($R = 0.72-0.77$; normal mean bias (NMB) = 3%-22%) and non-
23 episode days ($R = 0.88$; NMB = -21%), compared with the ceilometer at La Porte. Land-water
24 differences in PBL heights are captured better during non-episode days than episode days,
25 compared with the airborne High Spectral Resolution Lidar-2 (HSRL-2). During ozone episodes,
26 the simulated land-water differences are 50-60 m (morning), 320-520 m (noon), and 440-560 m
27 (afternoon) in comparison with the observed values of 190 m, 130 m, and 260 m, respectively.
28 During non-episode days, the simulated land-water differences are 140-220 m (morning) and 360-
29 760 m (noon) in comparison with the observed values of 210 m and 420 m, respectively. For
30 vertical ozone distributions, the model was evaluated against vertical profile measurements from
31 the Tropospheric Ozone lidar (TROPOZ), the HSRL-2, and ozonesondes, as well as at the surface
32 from a model 49i ozone analyzer and a site from the Continuous Ambient Monitoring Stations
33 (CAMS) at La Porte. The model underestimates free tropospheric ozone (2-3 km aloft) by 9%-22%
34 but overestimates near-ground ozone (< 50 m aloft) by 6%-39% during the two ozone episodes.
35 Boundary layer ozone (0.5-1 km aloft) is underestimated by 1%-11% during September 8-11 but
36 overestimated by 0%-7% during September 23-26. Based on these evaluations, we identified two
37 model limitations: the single-layer PBL representation and free tropospheric ozone
38 underestimation. These limitations have implications for the predictivity of ozone's vertical
39 mixing and distribution in other models.

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1 **1. Introduction**

2 The Houston metropolitan area has experienced nonattainment of the US National Ambient Air
3 Quality Standards (NAAQS) for ozone over decades (TCEQ, 2022). Ozone exceedances in
4 Houston usually occur in two peaks, a spring peak in April–May and a late summer peak in
5 August–October (Zhou et al., 2014). Such seasonal behavior is driven by diverse meteorological
6 conditions that influence ozone development. The passages of synoptic-scale cold fronts (~ 1000
7 km horizontally and ~ 5 km vertically; a timescale of days) are known to bring high background
8 ozone air from the continent into the Houston area (Lefer et al., 2010; McMillan et al., 2010;
9 Haman et al., 2014). Mesoscale sea breeze recirculation (~ 20 km horizontally and ~ 1 km
10 vertically; a timescale of hours) is found to be associated with ozone exceedances (Li et al.,
11 2020; Banta et al., 2005, 2011; Caicedo et al., 2019). Meanwhile, microscale-to-mesoscale
12 vertical mixing (< 1 km vertically; a timescale of hours) of the lower troposphere is shown to be
13 a significant factor in near-surface ozone air quality (Morris et al., 2010; Haman et al., 2014;
14 Sullivan et al., 2017; Xu et al., 2018; Caputi et al., 2019). Favored by these meteorological
15 conditions of different scales, local emissions of ozone precursors from the urban center and the
16 nearby Houston Ship Channel stay locally in the area and lead to high-ozone events. This study
17 will focus on the impact of mixing between lower free tropospheric layers on vertical ozone
18 distribution and the impact of chemistry is outside the scope of this analysis.

19 The planetary boundary layer (PBL) is the lower part (e.g. < 2 km) of the troposphere that is
20 directly influenced by the presence of the Earth’s surface and responds to surface forcings with a
21 timescale of an hour or less. A stable capping layer at the top of the PBL, where temperature
22 increases with height, is known as the capping inversion (CI) layer (e.g. ~ 2 km). With the cap in
23 place, air exchange is inhibited between the overlying free troposphere (FT) (e.g. >2 km) and the
24 underlying PBL (e.g. < 2 km). During the daytime, there is strong turbulence production
25 throughout the PBL, generating a buoyant layer called the convective boundary layer (CBL). The
26 CBL is characterized by intense mixing in a statically unstable situation where warm air rises
27 from the ground, growing from a few hundred meters in the early morning (e.g. ~ 0.5 km)
28 towards the top of the PBL in the afternoon (e.g. ~ 2 km). As the sun sets, convectively driven
29 turbulence decays in the formerly well-mixed CBL. The remnant of the recently decayed CBL
30 will remain aloft in the less-turbulent residual layer (RL) at around 1–2 km. As the night
31 progresses, the bottom portion of the RL transforms into a stable boundary layer (SBL) (e.g. <
32 0.5 km) due to its contact with the ground, characterized by statically stable air with weak and
33 sporadic turbulence. The PBL is commonly considered as the CBL under certain conditions
34 during the daytime and the SBL during the nighttime (Tangborn et al., 2021).

35 The heights of the PBL (incl. CBL and SBL) and other lower tropospheric layers (e.g. RL, CI)
36 are defined mainly by temperature inversions. It is primarily a thermodynamic-based definition.
37 Atmospheric models adopt the thermodynamic concept and rely on parameterization schemes to
38 define the structure of the PBL and compute the height of the PBL. Two major types of data
39 have been commonly used to validate the modeled PBL height. The first type is the PBL height
40 derived from the profiles of thermodynamic properties measured by ozonesondes and
41 radiosondes (Zhang et al., 2019; Zhang et al., 2020; Morris et al., 2010; Rappenglück et al.,
42 2008). These observations share a similar thermodynamic definition with the modeled PBL
43 height and are widely used to validate model prediction of the PBL height under various
44 conditions (day, night, land, water). The second type of data is remotely-sensed mixed layer



1 height as defined by aerosol backscatter gradients, which is becoming more widely available
2 with ceilometer data and aircraft lidars and can be adopted for model evaluation (Caicedo et al.,
3 2017, 2020; Knepp et al., 2017; Li et al., 2021; Wang et al., 2020).

4 Mixed layer height, defined as the volume of atmosphere in which aerosols are well mixed and
5 dispersed, can be derived from the unattenuated backscatter signal of aerosols alone (e.g. the
6 High Spectral Resolution Lidar-2 (HSRL-2)) or the attenuated total backscatter signal produced
7 by aerosols and molecules combined (e.g. CHM 15k-x ceilometers). Both signals have been used
8 to derive mixed layer height for model comparisons (Scarino et al., 2014; Li et al., 2022). Mixed
9 layer height does not equal PBL height by definition; it approximates the CBL height during the
10 daytime and can represent the height of the RL or the SBL depending on retrieval algorithms
11 applied to lidar signals at night (Wang et al., 2020; Vivone et al., 2021). Mixed layer height is
12 often a good proxy for the heights of different lower tropospheric layers determined
13 thermodynamically in models during the daytime (Scarino et al., 2014) and throughout the day
14 (Kuik et al., 2016; Haman et al., 2014) and serves as an input parameter of PBL heights for
15 meteorological and photochemical models (Tangborn et al., 2021; Knote et al., 2015; Geiß et al.,
16 2017).

17 Vertical mixing between different layers of the lower troposphere, such as boundary layer
18 mixing with the FT flow at its upper interface (through entrainment processes), mixing between
19 the RL and the SBL (through surface exchange processes) and the RL mixing through the growth
20 of the CBL, etc., strongly influences surface ozone concentrations. Entrainment can occur during
21 the daytime when strong convective thermals penetrate the laminar FT above and then sink back
22 into the CBL, bringing the FT air towards the surface and thus affecting surface ozone
23 concentrations (Parrish et al., 2010; Jaffe et al., 2011). Located between the FT and the CBL, the
24 strength of the CI layer limits the upward penetration of thermals and is thus used to indicate the
25 influence of the FT air on surface ozone (Kaser et al., 2017; Morris et al., 2010; Rappenglück et
26 al., 2008). Meanwhile, surface exchange processes occur when a low-level jet exists between the
27 RL and the underlying SBL and drives the shear production of turbulence between these layers.
28 Since the RL is a known ozone reservoir with limited NO_x titration and ozone deposition, ozone-
29 rich air in the RL can be mixed down into the SBL effectively, where it is subject to dry
30 deposition to the surface, affecting surface ozone concentrations (Tucker et al., 2010; Sullivan et
31 al., 2017; Caputi et al., 2019; Bernier et al., 2019; Zhao et al., 2022; Xu et al., 2018).

32 The Tracking Aerosol Convection Experiment Air Quality (TRACER-AQ, [https://www-](https://www-air.larc.nasa.gov/missions/tracer-aq/)
33 [air.larc.nasa.gov/missions/tracer-aq/](https://www-air.larc.nasa.gov/missions/tracer-aq/)) campaign, led by NASA with contributions from the Texas
34 Commission on Environmental Quality (TCEQ), probed Houston air quality with a
35 comprehensive suite of remote sensing and in situ measurements of ozone, ozone precursors, and
36 meteorology from ground-based, airborne, balloon-borne and shipborne platforms (Jensen et al.,
37 2022). The operational period occurred from July–September 2021, with intensive measurements
38 during September 2021. Combining field campaign observations with a high-resolution 3-D
39 photochemical model, the goals of this study are to (1) evaluate the PBL prediction in the model,
40 (2) examine the vertical distribution of ozone, and (3) identify specific model limitations that
41 prevent accurate prediction of the PBL height and the vertical ozone distribution.

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1 **2. Model and Data**

2 **2.1 Observations**

3 To evaluate the PBL and the vertical ozone distribution, this study adopted continuous, high-
4 resolution profiles (i.e., 1–10 minutes) from ground-based measurements at the La Porte site and
5 airborne measurements covering urban Houston and the Galveston Bay in September of 2021
6 (Fig. 1). Compared with discrete or low-resolution PBL measurements (*e.g.* hourly) used in
7 previous studies in Houston (Haman et al., 2014; Cuchiara et al., 2014; Rappenglück et al.,
8 2008), the high-resolution measurements in TRACER-AQ field campaign are capable to probe
9 into the fine PBL structure and its development as well as the associated vertical ozone profiles.

10 The La Porte site was equipped with (1) semi-continuous vertical ozone profiles from the NASA
11 Goddard Space Flight Center (GSFC) Tropospheric Ozone (TROPOZ) Differential Absorption
12 Lidar (DIAL) (Sullivan et al., 2014), (2) continuous aerosol mixed layer height derived from
13 atmospheric backscatter profiling with a CHM 15k-x ceilometer, (3) multiple ozonesonde
14 launches, and (4) continuous surface ozone and meteorology measurements.

15 The TROPOZ, as part of the ground-based Tropospheric Ozone Lidar Network (TOLNet,
16 <https://www-air.larc.nasa.gov/missions/TOLNet/>), has been used to provide continuous, high-
17 resolution profile measurements of vertical ozone profile during various campaigns for satellite
18 and model evaluation (Sullivan et al., 2014, 2015, 2019, 2022; Bernier et al., 2022; Kotsakis et
19 al., 2022; Dacic et al., 2020; Johnson et al., 2016; Dreessen et al., 2016). The TROPOZ data can
20 be used to identify pollutant transport to understand the vertical mixing of ozone. Similar to the
21 TROPOZ at the La Porte site, the University of Houston site measured semi-continuous vertical
22 ozone profiles with the Langley Mobile Ozone Lidar (LMOL) (Gronoff et al., 2019, 2021).

23 The CHM 15k-x ceilometer measured continuous atmospheric attenuated backscatter profiles at
24 a wavelength of 1064 nm. The signals were corrected due to the incomplete superposition of the
25 laser and the receiver field of view by the overlapping correction function from the manufacturer
26 (Rizza et al., 2017). The normalized range corrected signals (RCS) is shown in this paper. The
27 sharp gradients in the collected backscatter were then used to detect up to three aerosol layers by
28 the standard retrieval algorithm provided by the ceilometer manufacturer (Lufft, 2016). The
29 lowest determined aerosol layer is characterized as mixed layer height. It depends on the users to
30 determine whether the derived mixed layer height can be used as a proxy for thermodynamically-
31 defined layers such as the CBL, the SBL and the RL (Caicedo et al., 2017, 2020; Knepp et al.,
32 2017; Li et al., 2021; Wang et al., 2020).

33 Ozonesondes were often launched multiple times in a day at several locations and measured
34 vertical profiles of ozone and meteorological variables including temperature, humidity, and
35 winds. This study uses ozone and potential temperature profiles from eight ozonesondes at La
36 Porte launched on 10:00-15:00 CDT during ozone episodes.

37 Surface measurements at La Porte included ozone, air temperature, relative humidity, and wind
38 speed and direction. This study uses surface ozone from a Thermo Scientific model 49i ozone
39 analyzer operated by the GSFC and a TCEQ Continuous Ambient Monitoring Stations (CAMS)
40 site named La Porte Sylvan Beach, as well as surface meteorology from a Lufft WS-501B
41 operated by the GSFC.



1 In September 2021, the NASA Gulfstream-V aircraft flew on ten flight days. This analysis uses
2 the High Spectral Resolution Lidar-2 (HSRL-2) datasets collected over the Houston area up to
3 three times per day, roughly at 8:00-10:00, 11:00-13:00, and 14:00-16:00 local time, covering an
4 area of approximately 50 km x 135 km. With its high resolution and vertically resolved
5 measurements, the HSRL-2 demonstrated reliable performances on many previous airborne
6 campaigns (Hair et al., 2018; Hair et al., 2008; Burton et al., 2015). The HSRL-2 provides below
7 aircraft retrievals of the spatial and vertical distributions of ozone, aerosols, aerosol optical
8 properties, and mixed layer heights. This paper only reports on (1) mixed layer height derived
9 from gradients in the aerosol backscatter profiles measured at 532 nm and (2) ozone mixing ratio
10 along one flight track that has the nearest distance to the La Porte site (Fig. 1).

11 Mixed layer heights from the ceilometer at La Porte and the HSRL-2 are derived differently. The
12 ceilometer at the La Porte site measures attenuated total backscatter profiles of the atmosphere
13 (incl. aerosols and molecules), while the HSRL-2 can measure the unattenuated aerosol
14 backscatter profile. Both ceilometer and the HSRL-2 signals can be used to derive mixed layer
15 height. This study uses mixed layer heights from the ceilometer and the HSRL-2 to evaluate the
16 WRF-GC prediction of PBL heights.

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18 The supplement includes (1) surface measurements from the TCEQ CAMS and the boats
19 throughout July to October used to identify ozone exceedance days, (2) details on the
20 assimilation and evaluation of the modeled meteorology with these measurements, and (3)
21 vertical ozone distribution at University of Houston by LMOL.

22
23 Apart from the observations above, we used geopotential heights and winds at 850 hPa from the
24 European Centre for Medium Range Weather Forecast (ECMWF) reanalysis version5 (ERA5)
25 dataset (description in Sect. 2.3.2) to derive the synoptic conditions in Fig. 2.

26 27 **2.2 Identification of ozone episodes**

28 Ozone exceedance days were identified according to surface measurements from the TCEQ
29 CAMS (onshore) and the boats (offshore). The criteria used in this study is that any onshore site
30 from the CAMS network in Houston and Galveston or offshore boat ozone observations
31 registered daily maximum 8-hour average (MDA8) ozone in exceedance of 70 ppbv, the current
32 U.S. Environmental Protection Agency (EPA) NAAQS air quality standard for ozone. Three
33 high ozone episodes in September of 2021 were identified based on the above criteria:
34 September 6-11, September 17-19, and September 23-26, consisting of 13 ozone exceedance
35 days. We excluded analysis from the September 17-19 episode because it happened right after
36 tropical cyclone Nicholas, which made landfall 125 km south-southwest of Houston and
37 hindered measurements at the ground-sites and aircraft due to clouds and power outages. The
38 model meteorology was not designed to capture the cyclone either. Other September days were
39 used as a control to represent non-episode days.

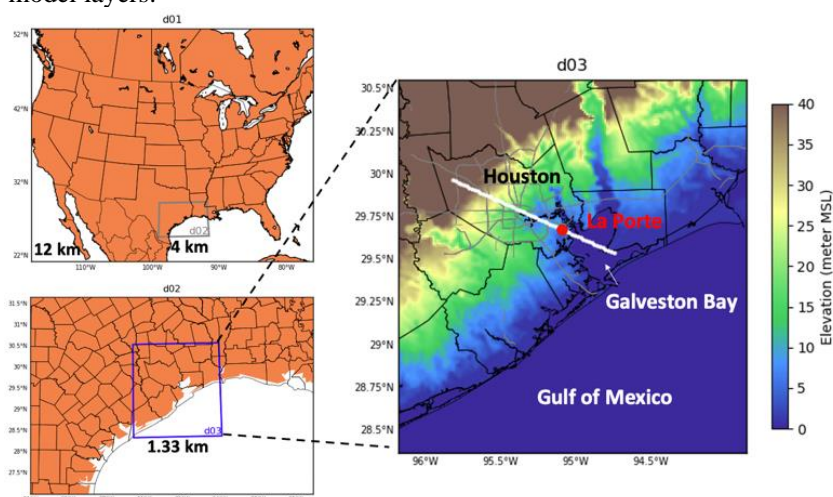
40 41 **2.3 Model**

42 **2.3.1 Model description**

43 WRF-GC v2.0 is a regional air quality model (Feng et al., 2021; Lin et al., 2020) that couples the
44 Weather Research and Forecasting (WRF) meteorological model (v3.9.1.1) with the GEOS-
45 Chem atmospheric chemistry model (v12.7.2). The WRF and GEOS-Chem versions are



1 benchmarks of WRF-GC v2.0 with the proven performance of meteorology, PBL heights, and
 2 aerosol simulation in Feng et al. (2021) and Lin et al. (2020). We evaluated WRF-GC's
 3 prediction of ozone during the TRACER-AQ study. We set up three domains with different
 4 horizontal resolutions that cover the contiguous United States, Southeast Texas, and the
 5 Houston-Galveston region, referred to as d01, d02, and d03, respectively, as shown in Figure 1.
 6 The corresponding horizontal resolutions for d01–d03 are 12 km, 4 km, and 1.33 km,
 7 respectively. All domains have identical vertical resolutions with 50 hybrid sigma-eta vertical
 8 levels spanning from the surface to 10 hPa. Vertical resolution ranges from ~70 m (near the
 9 ground) to ~700 m (aloft); the first 2 km above the ground has 10 model layers, and the first 4
 10 km has 14 model layers.



11
 12 Figure 1. WRF-GC nested domains and their horizontal resolutions. The La Porte site is labeled
 13 as a red dot. The white line represents a flight track that is chosen because of its nearest distance
 14 to the La Porte site.

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 16 WRF-GC uses the most updated full O_x - NO_x -VOC-halogen-aerosol chemistry from GEOS-
 17 Chem. The anthropogenic emissions used are the 2019 TCEQ emission inventory for Houston
 18 and southeastern Texas, the 2013 National Emission Inventory for the rest of the US, and the
 19 2014 Community Emissions Data System (CEDs) for regions outside of the US. Biomass
 20 burning emissions are from the 2019 Global Fire Emissions Database (GFED). Biogenic
 21 emissions are from the Model of Emissions of Gases and Aerosols from Nature (MEGAN)
 22 (Guenther et al., 2012). Soil NO_x (Hudman et al., 2012) and lightning NO_x (Murray et al., 2012)
 23 emissions are also included.

24 2.3.2 Model configurations

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 26 Boundary and initial conditions for WRF employed three alternative meteorological analyses.
 27 They were (1) the National Centers for Environmental Prediction (NCEP)-Final Analysis (FNL)
 28 (<https://rda.ucar.edu/datasets/ds083.3/>), (2) the fifth generation of European Centre for Medium-
 29 Range Weather Forecasts (ECMWF) atmospheric reanalysis (ERA5) data
 30 (<https://rda.ucar.edu/datasets/ds633.0/>), and (3) the High-Resolution Rapid Refresh (HRRR)
 31 from NOAA Amazon Web Service (<https://registry.opendata.aws/noaa-hrrr-pds/>). The temporal



1 resolution for FNL, ERA5, and HRRR is 6-hourly, hourly, and hourly, respectively. The
2 horizontal resolution for FNL, ERA5, and HRRR is 0.25°, 0.25°, and 3 km, respectively.
3
4 WRF has different schemes or options to represent physics and dynamics processes. Three PBL
5 schemes were used to investigate the effect of different parameterizations of heat, moisture, and
6 momentum exchange between the surface and PBL on the simulated PBL structure and height.
7 They are the local closure Mellor-Yamada-Nakanishi-Niino (MYNN) scheme (Nakanishi and
8 Niino, 2009), the non-local closure Yonsei University (YSU) scheme (Hong et al., 2006), and the
9 hybrid local-nonlocal Asymmetric Convective Model version 2 (ACM2) scheme (Pleim, 2007).
10 Details of the PBL schemes are in Sect. 2.3.3. Two microphysics schemes were used: the
11 Morrison double moment (2M) scheme (Morrison et al., 2009) and the single-moment 6-class
12 (WSM6) scheme (Hong and Lim, 2006). Other schemes adopted were the Monin-Obukhov
13 Similarity surface layer, the Noah land surface scheme (Chen and Dudhia, 2001), the Rapid
14 Radiative Transfer Model (RRTM) longwave and shortwave radiation schemes (Iacono et al.,
15 2008), and the New Tiedtke cumulus scheme (Zhang et al., 2011; Tiedtke, 1989).
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17 To select the best model configuration to represent meteorology during the 2021 TRACER-AQ
18 campaign, we designed eight model experiments with different physics options, boundary
19 meteorology, data assimilation, and reinitializing option, as listed in Table S2. First, [Base] is the
20 baseline configuration: MYNN for PBL, 2M for microphysics, NCEP FNL for boundary
21 conditions, no nudging for assimilation, and no reinitialization. Second, [YSU] and [ACM2]
22 experiments used the YSU and ACM2 PBL schemes, respectively, while keeping other options
23 the same as [Base]. Differences between [Base], [YSU], and [ACM2] show the effects of
24 different PBL parameterizations. Third, the [WSM6] experiment differs from [Base] by
25 replacing the 2M microphysics scheme with WSM6. Differences between [Base] and [WSM6]
26 show the effects of different microphysics schemes. Next, [ERA5] and [HRRR] were designed to
27 show the effects of different meteorological initial and boundary conditions on the WRF
28 performance by using ERA5 and HRRR instead of NCEP FNL, respectively. We examined the
29 effects of data assimilation options in [Nudged]. [Nudged] adopted observation nudging and
30 surface analysis nudging to assimilate both onshore and offshore measurements from multiple
31 platforms, including the TCEQ CAMS, boats, and the NCEP surface and upper air measurements
32 into WRF meteorology (see Text S2 for details). Differences between [Base] and [Nudged] show
33 the effects of assimilation. Last, [Reinit] used daily reinitialization where the simulation was
34 broken into many 30-hour segments with the first 6 hours of each segment (18:00-23:00
35 Coordinated Universal Time (UTC) of a previous day) as spin-up and the subsequent 24 hours
36 (0:00-23:00 UTC of the following day) used for analysis (Yahya et al., 2015; Otte et al., 2008).
37 Differences between [Base] and [Reinit] show the effects of a free-running option versus model
38 reinitialization.
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40 Among the above simulations, we chose four simulations (Table 1), including the three
41 simulations with different PBL schemes and the best simulation [HRRR] determined by
42 campaign-wide statistics (see Text S3 for details) in the analysis below. The surface layer, land
43 surface, longwave and shortwave radiation, and Tiedtke cumulus schemes remain unchanged in
44 all simulations.
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Table 1. List of simulations used in this study.

Simulations	Meteorology for Boundary Condition	PBL scheme
[Base]	NCEP FNL	MYNN
[YSU]	NCEP FNL	YSU
[ACM2]	NCEP FNL	ACM2
[HRRR]	HRRR	MYNN

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2.3.3 Determination of PBL height in different schemes

The heights of the PBL are determined differently among different PBL schemes in the WRF model. The intra-scheme differences can originate from (1) the vertical profile of thermodynamic quantities simulated with different assumptions of the vertical exchange of heat, moisture, and momentum and (2) the diagnosis of the PBL height from these thermodynamic quantities. The PBL heights determined by different schemes can differ by 20-30% (Hu et al., 2010; Xie et al., 2013).

First, the common parameterizations of vertical exchange include local and non-local closure schemes. Local closure schemes estimate the turbulent fluxes at each point in model grids from the mean atmospheric variables and their gradients at that point. In contrast, non-local closure schemes include the nonlocal upward transport by buoyant plumes, representing large-scale motions. Among the three PBL schemes used in this study, the MNYY scheme is local, the YSU is nonlocal, and the ACM2 is hybrid local-nonlocal.

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Second, the bulk Richardson number (BRN) and the turbulent kinetic energy (TKE) methods are the two common methods to determine PBL height. The BRN method diagnoses PBL height thermodynamically by potential temperature with wind speeds and is adopted in the YSU and the ACM2 schemes. The PBL heights under this condition are defined as the height of the model layer where the bulk Richardson number reaches a critical value. The two schemes have two major differences. The YSU scheme calculates the bulk Richardson number starting from the surface while the ACM2 scheme calculates it above the neutral buoyancy level (Hu et al., 2010; Hong et al., 2006; Pleim, 2007). The critical value is 0.25 for stable conditions and 0 for unstable conditions in the YSU scheme and it is 0.25 for both stable and unstable conditions in the ACM2 scheme (Xie et al., 2013). Meanwhile, the TKE method diagnoses PBL height by horizontal and vertical winds and is adopted in the MYJ scheme (not used in this study). The PBL height under this condition is diagnosed when the TKE decreases to a minimum of $0.1 \text{ m}^2 \text{ s}^{-2}$. A hybrid definition that combines the BRN and the TKE methods is implemented in the MYNN scheme. The hybrid method weights the TKE method more during stable conditions when the BRN-based PBL height is below $\sim 0.5 \text{ km}$, while it weights the TKE-based definition negligible when the BRN-based PBL height is above $\sim 1 \text{ km}$.

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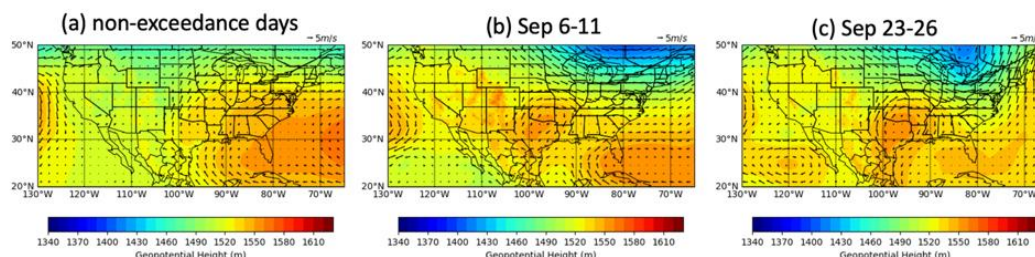
Previous studies have demonstrated that the mentioned schemes outperform each other under different conditions across regions, evaluated with various metrics (Hu et al., 2010; Xie et al., 2012; Xie et al., 2013). No conclusion is reached as to which scheme is universally the best. No systematic higher or lower PBL height is expected from one scheme relative to one another.



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3. Lower tropospheric layering for ozone-episode and non-ozone-episode days

The geopotential heights at 850 hPa in Figure 2 show different synoptic conditions are seen between ozone-episode and non-episode days in September 2021. The non-episode days experienced clean southerlies from the Gulf of Mexico (Fig. 2a), while the ozone episodes of September 6-11 and 23-26 both happened after a cold frontal passage with a low pressure sitting in the northeast US and a high pressure located in eastern Texas (Fig. 2b, 2c). This synoptic structure puts the Houston region under northerly wind conditions, which bring colder and more polluted continental air to the region, leading to relatively lower temperature (Fig. 3b) and relative humidity (Fig. 3d) than non-episode days (Fig. 3a; Fig. 3c).



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Figure 2. Synoptic conditions denoted by geopotential height at 850 hPa and the associated winds for (a) the non-episode days and the two ozone episodes of (b) September 6-11 and (c) September 23-26, 2021.

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Apart from differences in meteorological variables, synoptic high-pressure centers during ozone episodes tend to create a clear, calm condition with light horizontal winds at night when the RL and the multiple layer structure of the lower troposphere (incl. an SBL, an RL, and a CI layer) are prone to form, while the RL structure tends to be disrupted due to shear effects under meteorological conditions during non-episode days (Stull et al., 1988; Yi et al., 2001). We find mixed layer heights derived from the ceilometer at La Porte during non-episode days (Fig. 3i) and ozone-episode (Fig. 3j) are similar during the daytime, while the nocturnal mixed layer heights (*e.g.* 0:00–10:00 CDT) are greater on ozone-episode days than on non-episode days. Such differences can also be seen from the direct measurements of the ceilometer, *i.e.* atmospheric backscatter profiles, in Figures 3k and 3l. During ozone episodes, the high-pressure center creates favorable meteorological condition for multiple nocturnal layers to form. Among these, the RL contains much of the aerosol remnant left by the daytime CBL and is therefore detected by the ceilometer during ozone episodes (Fig. 3l). In contrast, no such multiple layers form under meteorological conditions on non-episode days. Much of the aerosol remnant above the SBL is dissipated with the disruption of RL by wind shear such that the SBL contains more aerosol than above. Therefore, the ceilometer detects the SBL on non-episode days (Fig. 3k). In this study, the mixed layer heights derived from the ceilometer represent the RL during ozone episodes but the SBL during non-episode days.

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Mixed layer height is often a good proxy for the heights of different lower tropospheric layers determined thermodynamically in models (Scarino et al., 2014; Kuik et al., 2016; Haman et al., 2014). We refer to the standard mixed layer retrievals, that is the CBL during the daytime, the SBL at night during non-episode days, and the RL at night during ozone episodes, respectively



1 as observed CBL, SBL or RL hereafter in a manner consistent with the modeled equivalents. The
2 next section evaluates the observed and the modeled PBL heights.

3

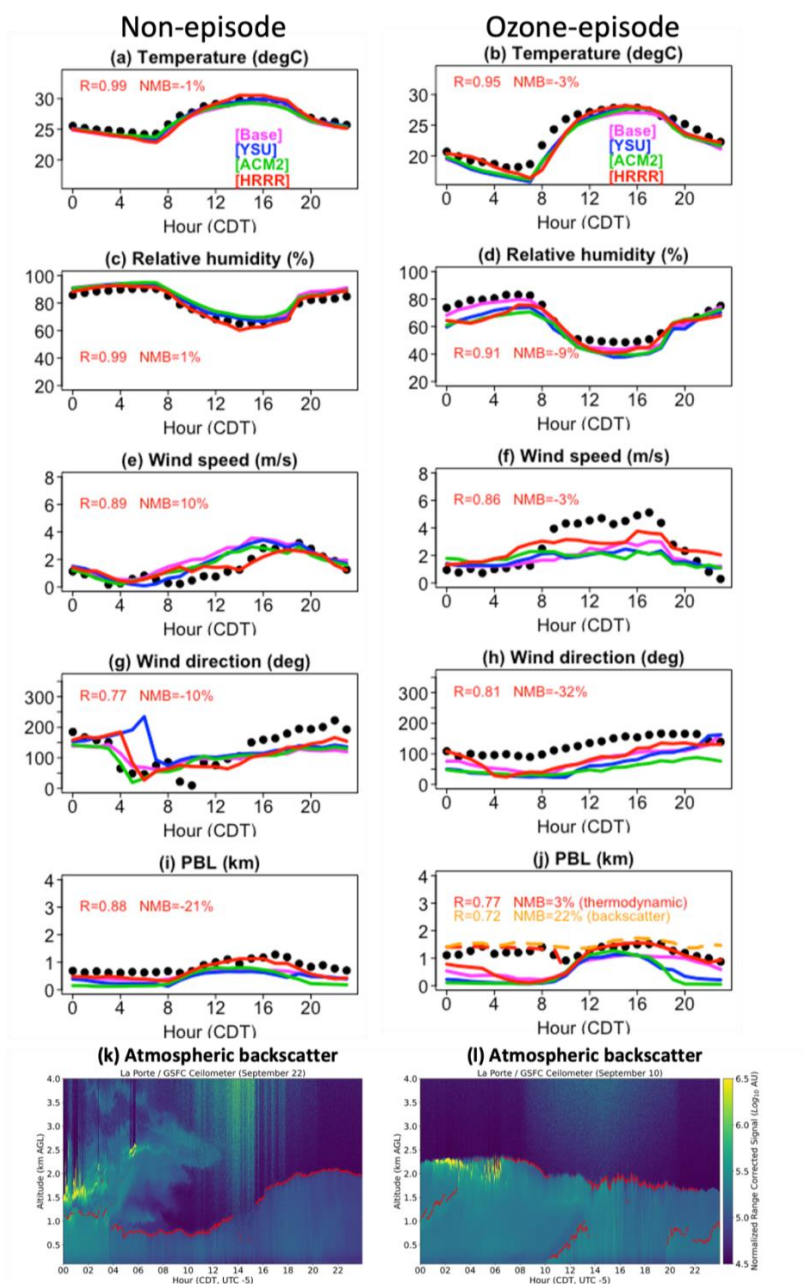
4 **4. PBL height evaluation**

5 In this section, we evaluate the modeled PBL height with two types of independent field
6 measurements. The ground-based ceilometer at the La Porte site is used to evaluate the diurnal
7 variation, given its continuous measurements throughout the day. Meanwhile, the HSRL-2
8 instrument acquired data over much of the urban Houston region and adjacent waters and is thus
9 used to evaluate spatial and temporal (daytime) variations of the PBL.

10

11 **4.1 Evaluation with ceilometer**

12 Diurnal variations of the PBL heights averaged during non-episode periods and ozone episodes
13 are separately evaluated in Figure 3. The observations represent the daytime CBL on both types
14 of days. At night, the observations represent the SBL on non-episode days but the RL on ozone-
15 episode days. The modeled equivalents are needed to yield equal comparisons between the
16 models and observations. The model diagnoses the CBL height as a standard output for the PBL
17 height during the daytime. However, the model only diagnoses the SBL as the standard output
18 for nighttime PBL rather than other nocturnal layers such as RL (Fig. 3j). Therefore, the modeled
19 RL needs to be extracted for a valid comparison with the observed RL during ozone episodes.



1
 2 Figure 3. Diurnal variation in observed versus modeled surface meteorology and PBL height
 3 averaged over different days during non-episode periods (left) and ozone episodes (right) in
 4 September 2021. For the first five rows, black dots are NASA GSFC observations at the La Porte
 5 site while lines are equivalent data simulated by the WRF-GC model. Different colors denote
 6 different model configurations. In (j), dashed lines represent residual layers identified by aerosol
 7 backscatter and potential temperature from the [HRRR] configuration. The last row shows the



1 ceilometer-measured atmospheric backscatter profiles overlaid with mixed layer heights of two
2 typical days; non-episode day September 22 (k) and ozone-episode day September 10 (l).
3

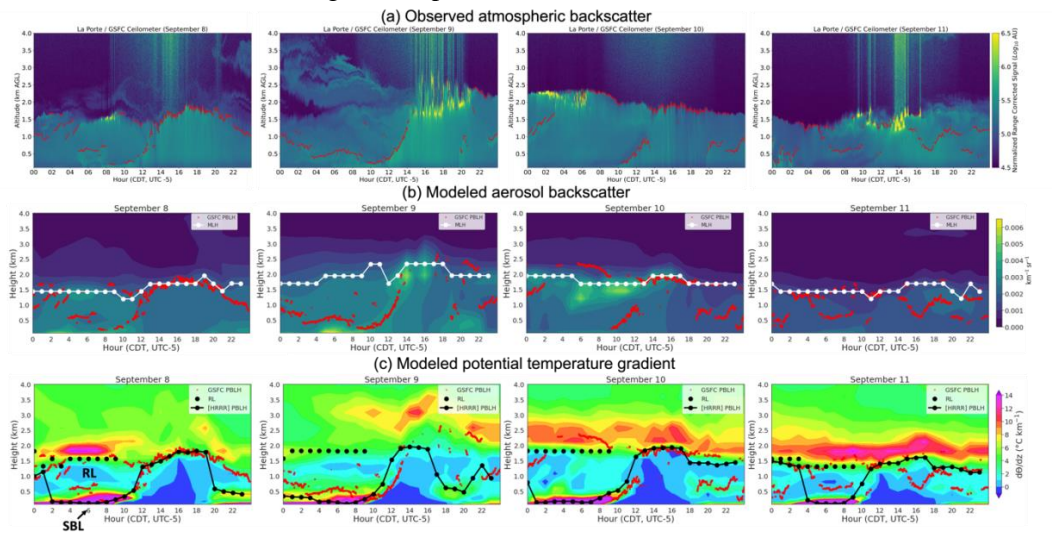
4 Before extracting RL characteristics in the model, we first selected one simulation with the best
5 daytime performance among the four simulations and examine its nighttime performance. On
6 non-episode days (Fig. 3i), [Base], [YSU], [ACM2], and [HRRR] respectively show the diurnal
7 mean and standard deviation of the PBL height of 0.52 ± 0.14 km, 0.43 ± 0.17 km, 0.39 ± 0.27 km,
8 and 0.66 ± 0.28 km in comparison with the observed value of 0.83 ± 0.22 km. On ozone-episode
9 days (Fig. 3j), the model simulations show the CBL height variation of 0.96 ± 0.18 km ([Base]),
10 0.60 ± 0.37 km ([YSU]), 0.50 ± 0.5 km ([ACM2]), and 1.25 ± 0.29 km ([HRRR]) in comparison
11 with the observed value of 1.26 ± 0.24 km during the afternoon and evening hours (15:00-23:00
12 CDT). During the same period, the model simulations show the PBL decay rates of 53 m h^{-1}
13 ([Base]), 102 m h^{-1} ([YSU]), 135 m h^{-1} ([ACM2]), 59 m h^{-1} ([HRRR]) in comparison with the
14 observed 60 m h^{-1} . All model simulations generally underestimate the PBL: 180–450 m
15 throughout the day on non-episode days and 10–760 m during the daytime on ozone-episode
16 days. The model underestimations are relative to the observed mixed layer height. The actual
17 PBL biases in the model can be larger or smaller than those underestimations depending on the
18 relationship between the backscatter-defined mixed layer and the thermodynamically defined
19 CBL. Among the four simulations, the [HRRR] best captures the observed mean height and
20 decay rate during the daytime and is thus examined further for the nighttime hours.
21

22 Second, the simulated aerosol backscatter (Fig. 4b, 5b) and potential lapse rates (Fig. 4c, 5c) of
23 the [HRRR] simulation are used to extract the modeled RL heights. The modeled aerosol
24 backscatter shows the volume of the atmosphere in which aerosol species are mixed and
25 dispersed. Substantially stronger backscatter signals are found within the first ~2 km than the
26 free troposphere at 3-4 km aloft with background backscatter of 0.01 - 0.02 km^{-1} sr^{-1} . Therefore,
27 we take the sharpest vertical gradient in the backscatter signal (i.e. the largest first derivative of
28 backscatter) to estimate the modeled mixed layer height. The extracted layers have daytime
29 variations of 1.58 ± 0.13 km and nighttime variations of 1.50 ± 0.06 km during ozone episodes. The
30 modeled aerosol backscatter in Figures 4b and 5b is not equivalent to the ceilometer-measured
31 atmospheric backscatter, which includes both aerosol and molecular backscatter signals in
32 Figures 4a and 5a. The modeled aerosol backscatter is presented here instead of the total
33 atmospheric backscatter because the latter is not diagnosed by the model. The modeled aerosol
34 backscatter is the closest product from the model to denote the modeled mixed layer heights.
35

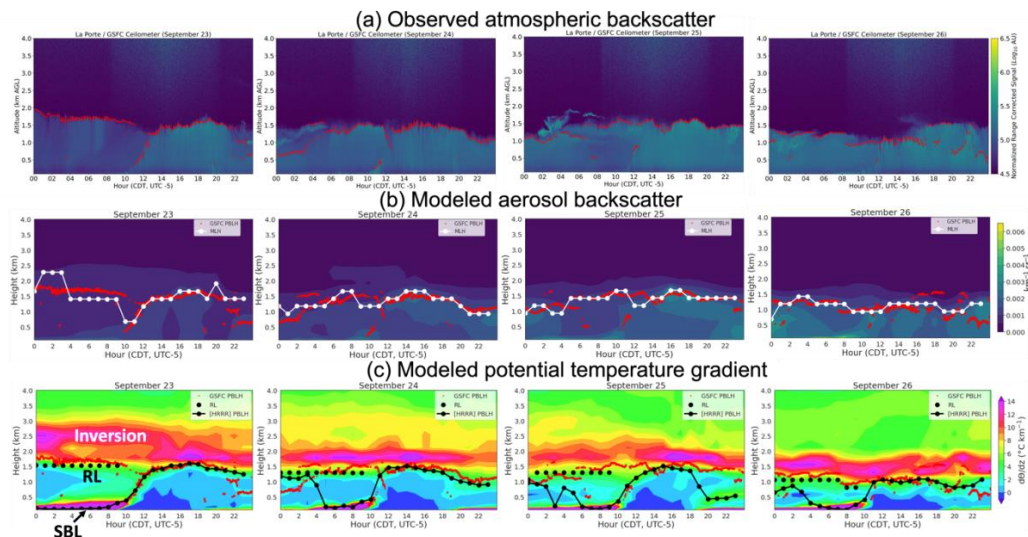
36 Potential lapse rate or potential temperature gradient $\left(\frac{d\theta}{dz}\right)$, defined thermodynamically as the
37 changes of potential temperature (θ) with height (z), is commonly used to distinguish
38 atmospheric layers according to their instability. Figures 4c and 5c show that the modeled
39 nocturnal PBL consists of a stable SBL, a neutrally stratified RL, and a CI layer during most
40 ozone-episode days. The modeled RL top is identified from where the RL (little or low
41 temperature increases at 0 - 3 $^{\circ}\text{C km}^{-1}$) shifts to the CI layer (drastic temperature increases at 8 -
42 14 $^{\circ}\text{C km}^{-1}$). Therefore, it can be identified by the sharpest gradient in the potential lapse rate,
43 which is 6.6 $^{\circ}\text{C km}^{-1}$ on average. The modeled RL top identified here has a variation of
44 1.39 ± 0.03 km during ozone episodes, slightly lower than the 1.50 ± 0.06 km identified by
45 backscatter.
46



1 The observed RL validates the backscatter-identified and thermodynamically-identified layers
2 from the model in Figure 3j. Model results have a slightly better agreement with the ceilometer
3 defined MLH for the thermodynamically-identified layer, with a correlation coefficient (R) of
4 0.77 and normalized mean bias (NMB) of 3%, than for the backscatter-identified layer, with
5 $R=0.72$ and $NMB=22\%$, during ozone episodes.



6
7 Figure 4. Observed and modeled heights of lower tropospheric layers at the La Porte site during
8 September 8-11. The contours show (a) ceilometer-observed attenuated atmospheric backscatter
9 signal produced by aerosols and molecules combined at 1064 nm, (b) modeled unattenuated
10 backscatter of aerosols alone at 1000 nm, and (c) modeled potential temperature gradient. Red
11 dots are ceilometer-observed mixed layer. White and black lines are backscatter-defined and
12 thermodynamically-defined mixed layers from the [HRRR] model simulation.



13
14 Figure 5. Same as Figure 4 but for September 23-26.



1

2 **4.2 Evaluation with HSRL-2**

3 This section evaluates spatial and temporal (daytime) variations of the modeled PBL heights
4 with the HSRL-2 mixed layer heights. The HSRL-2 collected measurements over the Houston
5 region and the adjacent Galveston Bay typically three times per day for ten days in September
6 2021. As stated in sections above, the mixed layer can represent the PBL under different
7 conditions and we refer to the observed mixed layer heights as observed PBL hereafter in a
8 manner consistent with the modeled equivalents.

9

10 The four model simulations underestimate the observed PBL heights under most conditions.
11 During ozone episodes, the urban Houston region observes the PBL heights of 0.63 ± 0.25 km in
12 the morning (8:00-10:00 CDT), 1.27 ± 0.38 km at noon (11:00-13:00 CDT), and 1.69 ± 0.23 km in
13 the afternoon (14:00-16:00 CDT). The observed heights are consistently lower over water with
14 values of 0.44 ± 0.34 km, 1.14 ± 0.45 km, and 1.43 ± 0.24 km for the three time periods,
15 respectively. Compared to observations, the four model simulations underestimate the heights
16 both over land (by 0.24-0.39 km in the morning, 0.02-0.25 km at noon, and 0.06-0.37 km in the
17 afternoon) and over water (by 0.02-0.25 km in the morning, 0.23-0.59 km at noon and 0.30-0.60
18 km in the afternoon). During non-episode days, the observed PBL height over land is 0.78 ± 0.14
19 km in the morning and 1.07 ± 0.24 km at noon. The morning heights are underestimated by 0.10-
20 0.34 km in the four model simulations, while the noon heights can be either underestimated by
21 0.25-0.37 km or overestimated by 0.05-0.23 km. In contrast to land, the observed height over
22 water is consistently lower with values of 0.57 ± 0.28 km in the morning and 0.65 ± 0.34 km at
23 noon; these are underestimated in the model by 0.03-0.28 km and 0.11-0.21 km, respectively.
24 Among the four simulations, the [HRRR] is the best to reproduce observed values with the
25 minimum model-observation differences under most conditions (e.g. different times and
26 regions), as shown in Table 2.

27

28 The observed land-water differences in PBL heights are larger in the afternoon than in the
29 morning during both ozone-episode and non-episode days. The four model simulations capture
30 the land-water differences better on non-episode days than ozone-episode days. During ozone
31 episodes, the observed mean land-water difference (land minus water) is 0.19 km while the
32 model predicts smaller differences of $-0.05\sim 0.06$ km in the morning; this is because the model
33 shows consistent larger underestimations over land than water. During noon and afternoon hours,
34 the observed mean land-water differences of 0.13 km and 0.26 km are predicted to be higher in
35 the model with values of 0.32-0.52 km and 0.44-0.56 km, respectively; this is because the model
36 shows consistent smaller underestimations over land than water during this period. During non-
37 episode days, the observed land-water gradients of 0.21 km (morning) and 0.42 km (noon) are
38 simulated to be 0.14-0.22 km and 0.36-0.76 km, respectively. The [ACM2] and the [HRRR]
39 slightly outperform the other two simulations for land-water differences (Table 2).

40

41 One ozone-episode day, September 9, is selected to show the spatial characteristics of different
42 simulations (Fig. 6). The four simulations match the observed mixed layer heights with high
43 spatial correlation at noon ($R=0.62\text{-}0.77$) and in the afternoon ($R=0.71\text{-}0.82$). Among all
44 simulations, the [HRRR] and the [Base] show the best spatial correlations at noon and in the
45 afternoon, respectively. The [HRRR] shows sharp transitions for the different flight legs over
46 urban Houston at noon on September 9 because of drastic changes in the modeled PBL heights at



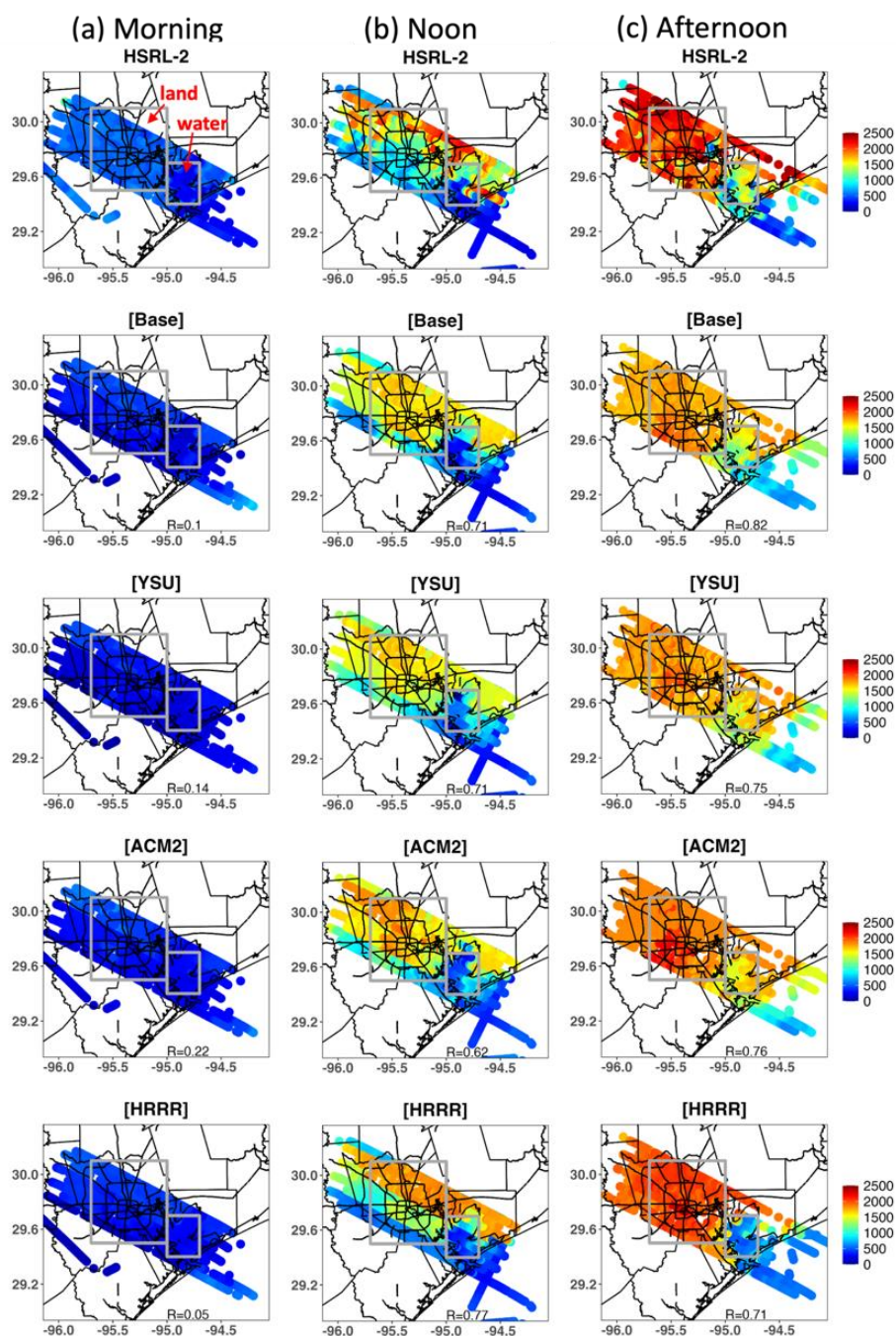
1 an hourly interval. The morning mixed layer heights can be difficult to retrieve with the
 2 influences from multiple layers (e.g. SBL and RL), and they can differ substantially from the
 3 thermodynamically-defined PBL. Therefore, we do not expect the model to capture the spatial
 4 patterns of mixed layer heights in the morning.

5
6

7 Table 2. Differences of the HSRL-2 mixed layer height and the WRF-GC thermodynamic PBL
 8 height during ozone-episode days (September 8-10 and 23-26) and non-episode days (September
 9 1 and 3). Land and water are defined by the gray boxes in Figure 6. The bias difference and the
 10 root mean square (RMS) difference are calculated by model simulations minus HSRL-2, with the
 11 unit of kilometers.

	Simulations	Differences	Morning (8:00-10:00 CDT)		Noon (11:00-13:00 CDT)		Afternoon (14:00-16:00 CDT)		
			Land	Water	Land	Water	Land	Water	
			Ozone-episode	[Base]	Bias	-0.242	-0.086	-0.240	-0.497
RMS	0.381	0.318			0.436	0.668	0.464	0.696	
[YSU]	Bias	-0.392		-0.250	-0.194	-0.589	-0.301	-0.610	
	RMS	0.488		0.409	0.406	0.785	0.409	0.807	
[ACM2]	Bias	-0.294		-0.167	-0.076	-0.471	-0.278	-0.457	
	RMS	0.430		0.378	0.376	0.683	0.476	0.665	
[HRRR]	Bias	-0.262		-0.026	-0.040	-0.232	-0.068	-0.303	
	RMS	0.384		0.312	0.289	0.462	0.223	0.455	
Non-episode	[Base]	Bias		-0.211	-0.218	-0.243	-0.276		
		RMS		0.309	0.353	0.433	0.440		
	[YSU]	Bias		-0.348	-0.282	-0.363	-0.304		
		RMS		0.434	0.397	0.506	0.471		
	[ACM2]	Bias	-0.236	-0.236	0.050	-0.261			
		RMS	0.371	0.377	0.460	0.440			
	[HRRR]	Bias	-0.100	-0.029	0.237	-0.107			
		RMS	0.243	0.301	0.377	0.364			

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Figure 6. Spatial variabilities of the PBL heights (in meters) from the HSRL-2 and different WRF-GC simulations (a) in the morning (8:00-10:00 CDT), (b) at noon (11:00-13:00 CDT), and (c) in the afternoon (14:00-16:00 CDT) of September 9, 2021.



1 **5. Ozone vertical mixing and distribution**

2 Boundary layer mixing can bring air aloft towards the surface and vice versa, leading to uneven
3 vertical distribution of ozone which accordingly affects surface ozone concentrations. This
4 section uses independent field measurements at La Porte (incl. TROPOZ, HSRL-2, ozonesondes,
5 a model 49i ozone analyzer, and a CAMS site named La Porte Sylvan Beach) to validate the
6 modeled vertical ozone profiles at three layers, including the lower free troposphere (2-3 km
7 aloft), the boundary layer (0.5-1 km aloft), and the ground level (<50 m). Since the [HRRR]
8 simulation best represents the PBL variations in Section 4, it is used to investigate vertical ozone
9 profiles in this section.

10

11 **5.1 Free tropospheric ozone entrainment**

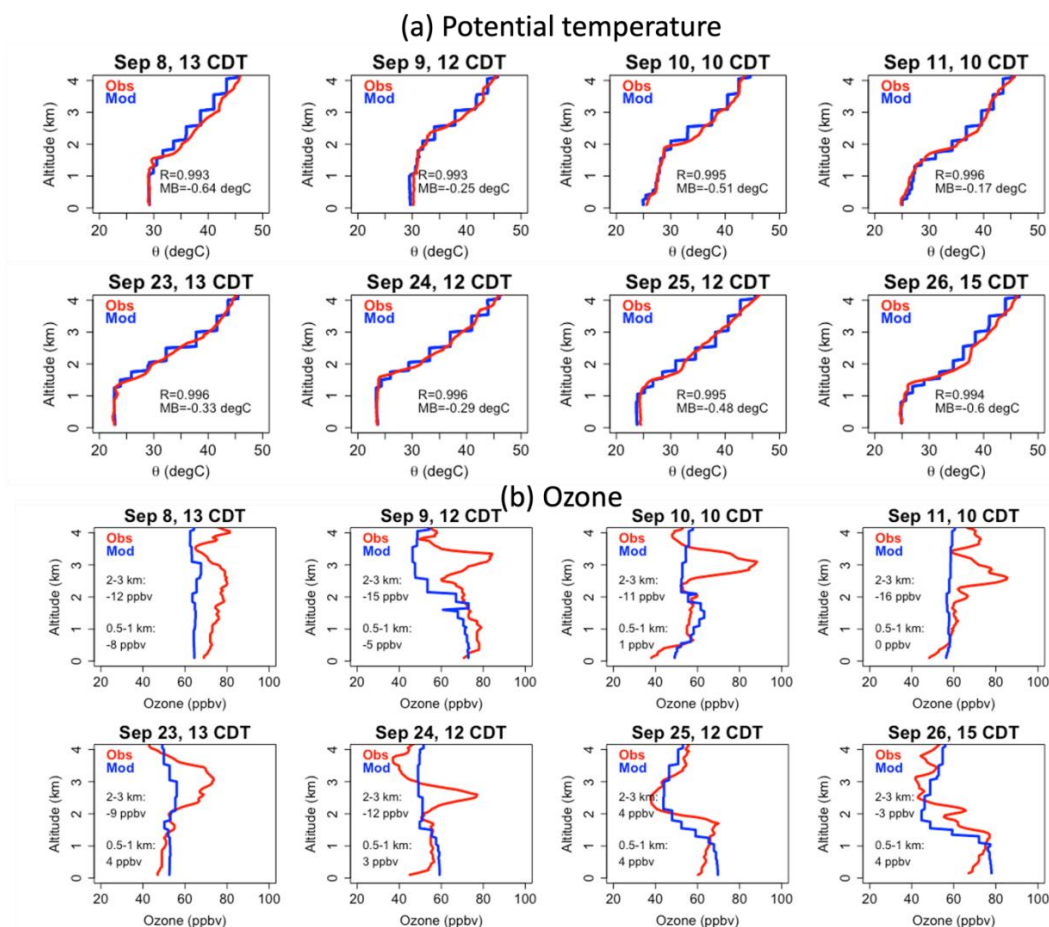
12 The strength of the CI layer regulates the gas exchange between the FT and the PBL. Strong
13 convection can penetrate a weak CI layer and entrain FT air into the PBL (i.e. entrainment),
14 while a strong CI layer acts as a lid to restrict gas exchange between the PBL and the FT. The
15 potential temperature differences between the top and bottom of the CI layer are often used to
16 indicate the strength of the CI layer and the extent of entrainment processes (Kaser et al., 2017;
17 Morris et al., 2010; Rappenglück et al., 2008). We first identified the modeled CI layers at 1.5–3
18 km aloft during ozone episodes (Fig. 4c and Fig. 5c), and then calculate the temperature
19 differences in the model between the top and bottom of the CI layers in each day. The
20 corresponding daily inversion strength is 2.3 °C, 2.8 °C, 6.8 °C and 6.4 °C during September 8-
21 11 and 13.6 °C, 7.5 °C, 7.8 °C, and 8.4 °C during September 23-26, respectively. Among these
22 days, September 8 and 9 experienced the weakest inversions. To examine if the modeled
23 inversion strength is representative of the observations, we evaluate the modeled potential
24 temperature profiles with ozonesonde measurements in Fig. 7a. Results show that the model
25 simulates the vertical profiles of potential temperature well across different days with high
26 correlation ($R=0.99$) and low biases ($MB= -0.64\text{ °C} \sim -0.17\text{ °C}$).

27

28 Combining the inversion strengths (Fig. 4c; Fig. 5c) and the vertical ozone distributions from the
29 TROPOZ lidar (Fig. 8) helps to identify potential entrainment of the FT air into the underlying
30 PBL on September 8 and 9 at the La Porte site. On September 8, strong convection associated
31 with a rapid CBL growth penetrates the thin and weak inversion at 2 km aloft at around noon
32 (Fig. 4c) and allows the ozone-rich air above to entrain into the CBL, adding to afternoon ozone
33 buildup (Fig. 8a). Similarly, there is no CI layer present overnight from 20 CDT on September 8
34 to 10 CDT on September 9 (Fig. 4c) and thus long-lasting ozone entrainment into the RL (Fig.
35 8a). Conversely, a strong and thick inversion at 1.5-3 km decouples the FT and the underlying
36 PBL during September 23-24 (Fig. 5c) and the ozone layer remains aloft at 2-3 km (Fig. 8d). The
37 inversion strength presented here is one way to approach the potential entrainment, follow-up
38 studies can probe into the detailed dynamics. It is also noteworthy that the presented vertical
39 distribution of ozone is also largely shaped by local ozone production in the boundary layer.
40 Since this study is focused on the vertical ozone distribution impacted by mixing between lower
41 free tropospheric layers, the vertical ozone distribution impacted by chemistry and differentiating
42 between the contributions from dynamics and chemistry are outside the scope of this analysis.

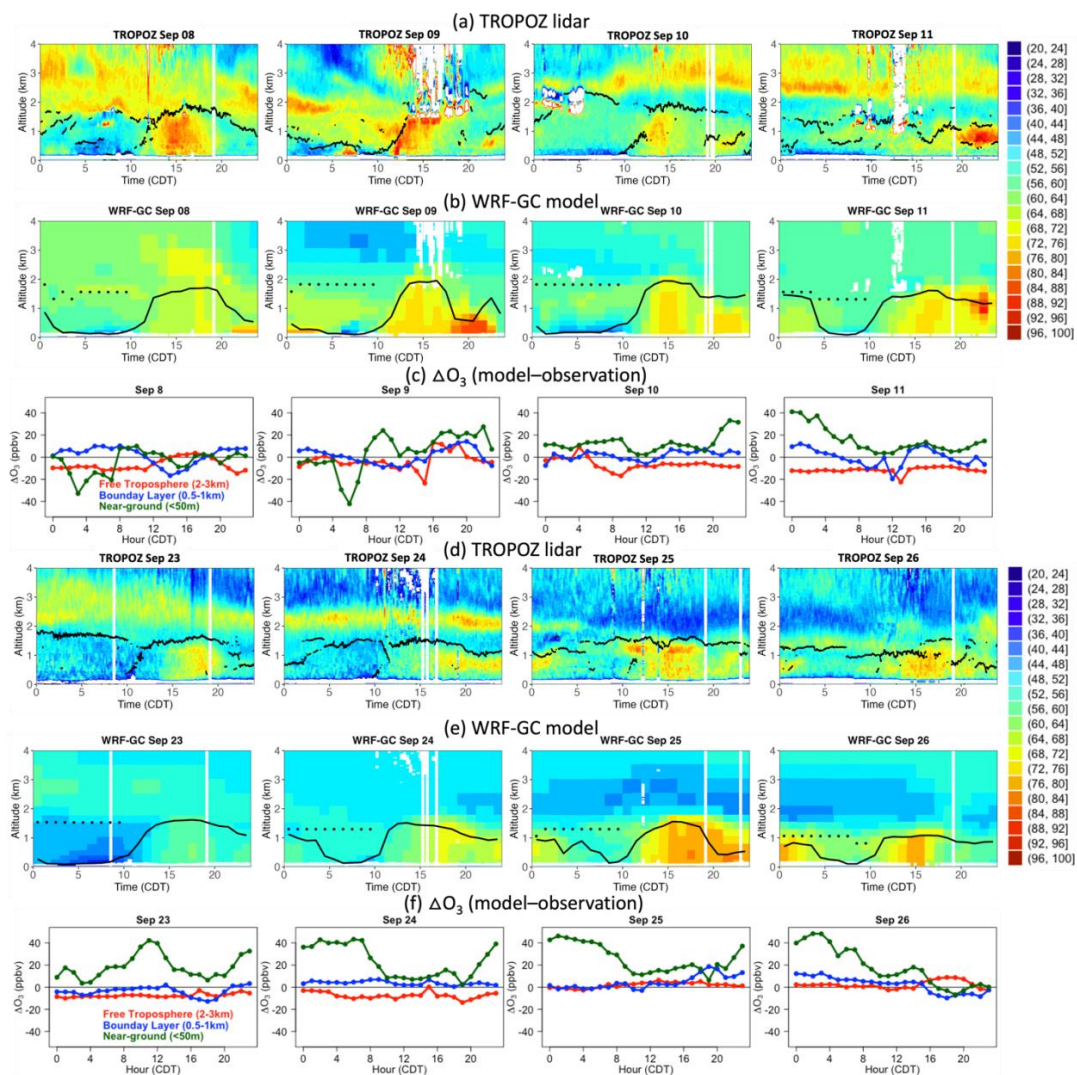
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1
2 Figure 7. Vertical profiles of (a) potential temperature and (b) ozone from ozonesonde
3 measurements and the WRF-GC [HRRR] simulation at La Porte during September 8-11 and
4 September 23-26.

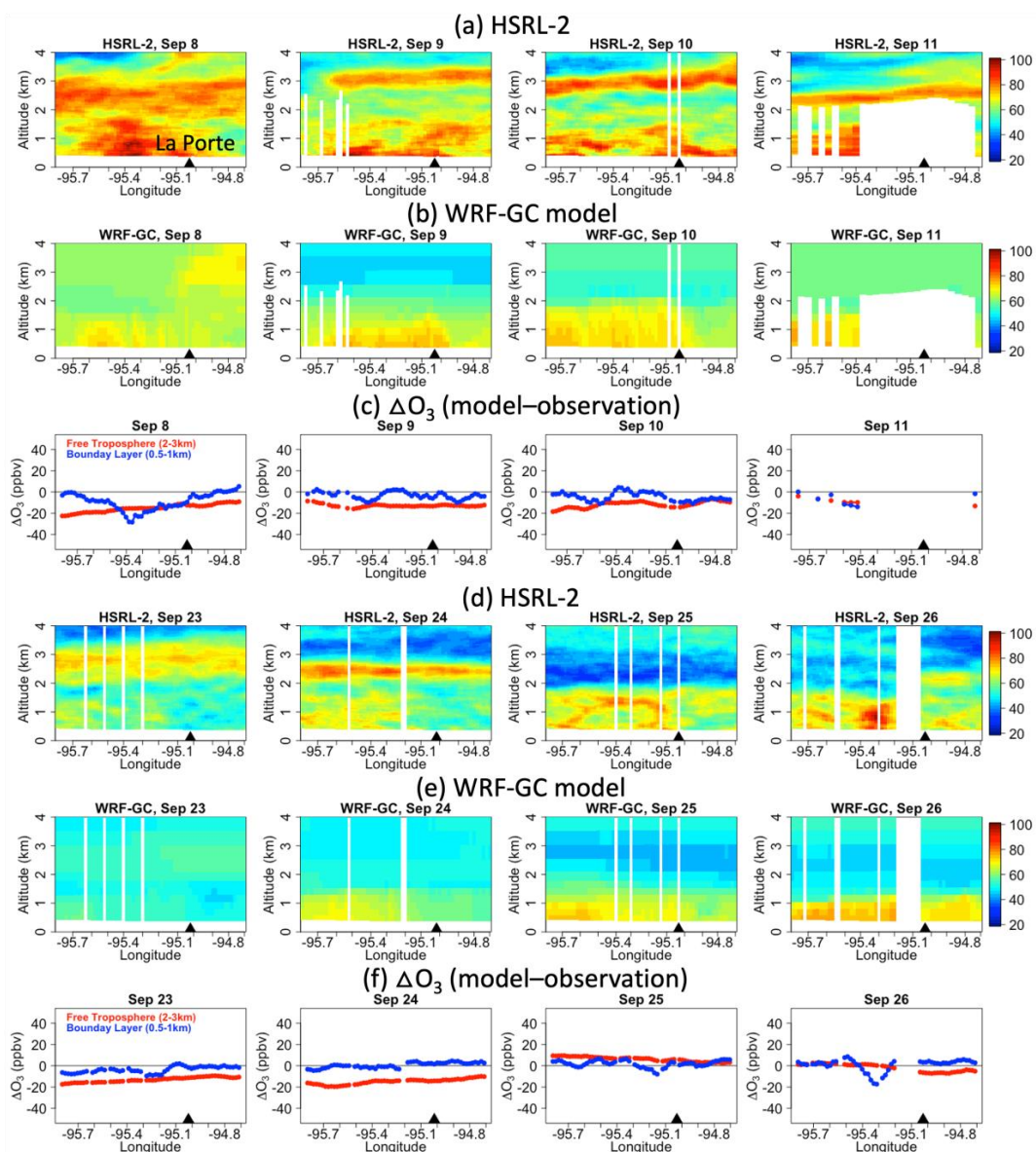
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 2 Figure 8. Time series of the vertical ozone profile from the TROPOZ ozone lidar (a, d) and the
 3 WRF-GC [HRRR] simulation (b, e) at La Porte. Observed and modeled boundary layer heights
 4 are inserted, respectively. Dots represent the modeled residual layer identified in this study. Line
 5 plots (c, f) show ozone differences (model minus observation) at the free troposphere (2-3 km)
 6 and the boundary layer (0.5-1 km) from the TROPOZ as well as the near-ground (< 50m) from
 7 the model 49i ozone analyzer.
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Figure 9. Vertical ozone profiles from (a, d) the HSRL-2 and (b, e) the WRF-GC [HRRR] simulation. The profiles are taken from a flight track (Fig.1) over urban Houston and Galveston Bay at around 11:00-13:00 CDT each day. Line plots (c, f) show ozone differences (model minus observation) at the free troposphere (2-3km) and the boundary layer (0.5-1km).



1 **5.2 Evaluation of ozone vertical distribution**

2 Multiple field measurements at La Porte are used to evaluate the modeled vertical ozone
3 distribution at the free troposphere (2-3 km), the boundary layer (0.5-1 km), and the near-ground
4 level (< 50m). According to data availabilities at different levels, the free troposphere and
5 boundary layer are evaluated by the TROPOZ, the HSRL-2, and ozonesondes (Table 3), while
6 the ground level is evaluated by a model 49i ozone analyzer and a CAMS site named La Porte
7 Sylvan Beach (Table 4). To cross compare among multiple measurements, we present the model-
8 observation differences at a common site (La Porte) during a common time slot (11:00-13:00
9 CDT). Larger ozone differences are found at the near-ground level than for the boundary layer
10 and lower free troposphere (Table 3; Table 4).

11
12 As shown in Table 3, the model underestimates the layer of enhanced ozone at 2-3 km aloft in
13 the free troposphere by 9%-21% (TROPOZ), 15%-22% (HSRL-2), and 14%-22% (ozonesondes)
14 at La Porte at 11:00-13:00 CDT during September 8-11 and September 23-24. Unlike most of the
15 campaign's ozone exceedance days, September 25 and 26 do not have an enhanced ozone layer
16 at 2-3 km aloft but have a lower ozone layer relative to the background tropospheric values
17 instead; this low ozone layer is overestimated by 9%-12% on September 25 but underestimated
18 by 3-12% on September 26. Meanwhile, the model underestimates the boundary layer ozone at
19 0.5-1 km aloft by 6%-10% (TROPOZ), 6%-12% (HSRL-2), and 1%-11% (ozonesondes) during
20 the first ozone episode of September 8-11, but overestimate it by 0%-6% (TROPOZ), 3%-5%
21 (HSRL-2), and 5%-7% (ozonesondes) during the second episode of September 23-26. Above
22 model-observation differences are based on the common site (La Porte) and common time
23 (11:00-13:00 CDT) among different measurements, the temporal (Figures 8c and 8f) and spatial
24 (Figures 9c and 9f) variations of these differences are shown in Figures 8 (TROPOZ) and 9
25 (HSRL-2).

26
27 While free tropospheric and boundary layer ozone are important components of the vertical
28 ozone distribution due to their thickness, the thin layer of near-ground ozone affects human and
29 vegetation health the most and thus receives more attention. In Table 4, the model overestimates
30 near-ground ozone by 6%-24% (model 49i ozone analyzer) and 8%-39% (CAMS La Porte
31 Sylvan Beach) at La Porte at 11:00-13:00 CDT during the two ozone episodes. Figures 8c and 8f
32 show the temporal variations of model-observation differences from the model 49i ozone
33 analyzer. Most near-ground ozone differences occur at night, consistent with the known problem
34 of overestimating nighttime ozone common to many photochemical models (Schnell et al., 2015;
35 Travis et al., 2016; Jaffe et al., 2018). The WRF-GC model adopts a chemical module from
36 GEOS-Chem. Thus, the two share the difficulties replicating nighttime ozone due to reasons
37 such as the insufficient representation of the stratification of multiple nocturnal atmospheric
38 layers, uncertainties in gas exchanges between the residual layer and the underlying surface
39 layer, and difficulties in simulating the timing of changes in PBL dynamics (Travis and Jacob,
40 2019).

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1 Table 3. Absolute (abs.) and percentage (pct.) ozone differences between field measurements and
 2 the model at free troposphere and boundary layer at La Porte.

		TROPOZ (11-13 CDT)		HSRL-2 (11-13 CDT)		Ozonesonde (10-15 CDT)	
		abs. (ppbv)	pct.	abs. (ppbv)	pct.	abs. (ppbv)	pct.
Free Troposphere (2-3km aloft)	8-Sep	-7	-9%	-12	-15%	-12	-16%
	9-Sep	-8	-13%	-13	-20%	-15	-22%
	10-Sep	-8	-13%	-14	-21%	-11	-18%
	11-Sep	-16	-21%			-16	-21%
	23-Sep	-8	-13%	-11	-17%	-9	-14%
	24-Sep	-9	-15%	-14	-22%	-12	-20%
	25-Sep	5	12%	6	15%	4	9%
	26-Sep	-2	-3%	-6	-12%	-3	-6%
Boundary Layer (0.5-1km aloft)	8-Sep	-5	-7%	-8	-11%	-8	-11%
	9-Sep	-8	-10%	-5	-6%	-5	-7%
	10-Sep	-4	-6%	-9	-12%	1	2%
	11-Sep	-5	-7%			-0.4	-1%
	23-Sep	0	0%	-2	-3%	4	7%
	24-Sep	2	4%	2	4%	3	5%
	25-Sep	1	2%	2	3%	4	7%
	26-Sep	4	6%	3	5%	4	5%

3
 4 Table 4. Absolute (abs.) and percentage (pct.) ozone differences between field measurements and
 5 the model at the near-ground level at La Porte.

		Model 49i (11-13 CDT)		CAMS La Porte Sylvan Beach (11-13 CDT)	
		abs. (ppbv)	pct.	abs. (ppbv)	pct.
Near-ground (< 50m)	8-Sep	4	7%		
	9-Sep	8	12%		
	10-Sep	4	6%	16	31%
	11-Sep	6	9%	15	29%
	23-Sep	4	9%	4	8%
	24-Sep	10	20%	15	34%
	25-Sep	13	24%	18	33%
	26-Sep	12	21%	21	39%

6



1 **6. Conclusion**

2 We used ground-based and aircraft observations collected during the TRACER-AQ campaign in
3 September 2021 to evaluate WRF-GC simulation of the PBL height and ozone in Houston,
4 including two ozone episodes characterized by MDA8 ozone exceeding 70 ppbv. The combined
5 suite of ground-based and airborne meteorological and chemical observations are critical in
6 thoroughly evaluating the spatial and temporal variations of the PBL and vertical ozone
7 distributions during multi-day ozone episodes, as presented in this work.

8
9 The modeled PBL heights are evaluated with mixed layer heights retrieved by the ground-based
10 ceilometer and the airborne HSRL-2. Compared with the ceilometer, the four model simulations
11 of [Base], [YSU], [ACM2], and [HRRR] generally underestimate the PBL heights to different
12 extents: 180–450 m throughout the day on non-episode days and 10–760 m during the daytime
13 on ozone-episode days. As the best simulation, the [HRRR] captures the diurnal variations
14 during non-episode days ($R=0.88$; $NMB=-21\%$). Standard models do not diagnose RL heights,
15 unlike ceilometers. Therefore, we separately identified the modeled RL following the practices
16 using aerosol backscatter signals and potential temperature gradients during the ozone episodes.
17 As a result, the diurnal variation of the thermodynamically-identified layer ($R=0.77$; $NMB=3\%$)
18 compares slightly better than that of the backscatter-identified layer ($R=0.72$; $NMB=22\%$)
19 during ozone episodes. Meanwhile, when compared with the HSRL-2, the four simulations
20 underestimates PBL heights by 20–390 m over the urban Houston region and by 20–600 m over
21 the adjacent Galveston Bay during ozone episodes. On non-episode days, the PBL heights over
22 urban region are either underestimated by 100–370 m or overestimated by 50–230 m and those
23 over the Bay are underestimated by 30–210 m by the four simulations. On both ozone-episode
24 and non-episode days, the observed land-water differences in PBL heights are larger in the
25 afternoon than in the morning; the model captures such daytime trends. The four model
26 simulations capture the land-water differences better on non-episode days than ozone-episode
27 days.

28
29 We evaluated the vertical ozone distribution with multiple field measurements, including
30 TROPOZ, HSRL-2, ozonesonde, a model 49i ozone analyzer, and a CAMS site named La Porte
31 Sylvan Beach. Evaluations were done at three lower tropospheric layers: the free troposphere (2–
32 3 km aloft), the boundary layer (0.5–1 km aloft), and the ground level (< 50 m aloft). As a result,
33 the model underestimates the high ozone layer in the free troposphere by 9%–21% (TROPOZ),
34 15%–22% (HSRL-2), and 14%–22% (ozonesondes) on most ozone-episode days. The boundary
35 layer ozone is underestimated by 6%–10% (TROPOZ), 6%–12% (HSRL-2), and 1%–11%
36 (ozonesondes) during September 8–11, but overestimated by 0%–6% (TROPOZ), 3%–5%
37 (HSRL-2), and 5%–7% (ozonesondes) during September 23–26. Meanwhile, the model
38 overestimates near-ground ozone by 6%–24% (model 49i ozone analyzer) and 8%–39% (CAMS
39 La Porte Sylvan Beach) during the two ozone episodes.

40
41 Based on these evaluations, we summarized model limitations that prevent more accurate
42 simulation of PBL heights and the vertical ozone distribution during TRACER-AQ. The first
43 limitation is the single-layer PBL representation. The WRF model only diagnoses the SBL at
44 night, despite the model simulating different physical and thermodynamic properties of multiple
45 nocturnal layers above the SBL. For example, the RL is not identified by the model as a standard
46 diagnosis; this prevents the direct comparison of the model outputs with the observed RL at



1 night. The second limitation is the underestimation of the layer of enhanced ozone 2-3 km aloft
2 in the free troposphere that was often present on ozone-episode days during the campaign. Given
3 its height of 2-3 km and a lifetime of around a week, the layer of enhanced ozone was likely
4 transported into Houston by synoptic flows of cold fronts from the north. The
5 underrepresentation of the synoptic layer of enhanced ozone affects model representations across
6 regions horizontally and atmospheric layers vertically, making it particularly important to model
7 vertical ozone distributions and the effects of entrainment accurately.

8
9 Our findings of the model limitations have implications for the predictivity of ozone's vertical
10 mixing and distribution across different modeling systems. For example, WRF is widely used in
11 various meteorology-chemistry coupling systems with different treatments of boundary layer
12 mixing. In WRF-Chem, boundary layer mixing in the chemistry part uses a mixing coefficient
13 originating in WRF such that the boundary layer mixing calculations in the meteorology and
14 chemistry parts share the same set of coefficients. In WRF-GC, the chemistry part from GEOS-
15 Chem only takes the PBL height from WRF as the maximum height for boundary layer mixing
16 and conducts independent calculations of boundary layer mixing from WRF. Unlike online
17 coupled WRF-Chem and WRF-GC, WRF is offline coupled to CAMx in the WRF-CAMx
18 system, and the boundary layer mixing in the chemistry part of CAMx is subject to WRF output
19 frequency instead of the native transport time step in WRF. Thus, it is essential to understand
20 how the model limitation of a single-layer PBL representation affects boundary layer mixing in
21 chemical simulations among different meteorology-chemistry coupling systems. Follow-up
22 studies to this work will address these aspects with detailed model intercomparisons.

23
24 **Code availability.** WRF-GC is a free and open source model (<http://wrf.geos-chem.org>; last
25 access: 29 May 2023) (Lin et al., 2019; Feng et al., 2021). The two parent models, WRF and
26 GEOS-Chem, are also open source and can be obtained from their developers at
27 <https://github.com/wrf-model/WRF> (last access: 29 May 2023) and <http://www.geos-chem.org>
28 (last access: 29 May 2023), respectively. The version of WRF-GC (v2.0) described in this paper
29 couples WRF v3.9.1.1 and GEOS-Chem v12.7.2 and is archived in Zenodo at
30 <https://doi.org/10.5281/zenodo.4395258> (last access: 29 May 2023).

31 **Data availability.** All observation datasets, model configuration files, model boundary
32 conditions, model input files, and scripts used in this paper are archived in Zenodo at
33 <https://doi.org/10.5281/zenodo.7983449> (last access: 29 May 2023).

34
35 **Author contributions.** XL and YW conceived the research idea. XL wrote the initial draft of the
36 paper and performed the analyses and model simulations. JF, TG, and SA provided the shipborne
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41
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11

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