Review of the manuscript titled "The impact of observation nudging on simulated meteorology and ozone concentrations during DISCOVER-AQ 2013 Texas campaign" By Xiangshang Li et al. Submitted to the ACP

Recommendation: minor revision

General response:

Thanks for the very helpful comments. We think the reviewer's comments are based on an earlier version (v1) of the manuscript. The version in discussion (ACPD) has added several missing references pointed out by the reviewer. It also has several new plots which we think are better than old version (please refer to uploaded updated manuscript according to the first reviewer's comments too). Also there are enough differences from v1 such that the line number changed substantially. In this response, we will refer to the version in ACPD regarding to line number through keeping the original line number in item title.

1. Comments Line 11-12, it is confusing to say observational nudging is objective analysis and to use "OA" representing observational nudging for the rest of the paper. In the WRF modeling community, objective analysis usually refers to "OBSGRID" which is an extra package in the WRF model system to take observations to improve the first guess fields (WPS output files) through an objective analysis method. Observational nudging is the Newtonian relaxation method implemented in WRF code to use observations to minimize model error during the simulation. It is known as "four-dimensional data assimilation (FDDA)" introduced by Stauffer and Seaman (1990). The observational nudging is not necessary to be run with objective analysis package even though the OBSGRID provides the observational data written in required by WRF for nudging. Stauffer, D., Seaman, N.L., 1990. Use of four-dimensional data assimilation in a limitedarea mesoscale model. Part I: Experiments with synoptic-scale data. Month. Weather Rev. 110, 1250e1277.

Reply: Thanks for pointing out the issue. We understand the differences between "objective analysis" and "observation nudging". In WRF, to perform obs-nudging, one has to use OBSGRID (Objective Analysis) to generate necessary input. As seen in the flow chart of OBSGRID (usually on page 7-2 of WRF-ARW User's Guide), after OBSGRID is run, three types of output files (metoa_em for OA, wrfsfdda for surface analysis nudging and OBS_DOMAIN for obs-nudging) are generated and all of them are to be used. This is because performing three tasks (OA+sfdda+obs-nudging) can maximize the benefit of assimilating observations. Therefore, running OBSGRID tends to imply performing all three tasks. Although one can do obs-nudging without performing OA (i.e., discard "metoa_em"), this should not be considered normal since it is likely degrade WRF performance.

To address the issue, we made two changes. We

1) included an explanation on objective analysis and obs-nudging in introduction (line 65-75) and clarified the case names in section 2.2

2) changed all the OA in paper to "obs-nudging" except for case names. Using "OA" and "No-OA" as case name seems easier.

2. Line 70-71, what does this mean??? There are detailed statistics about the nudging impacts shown in those studies (Deng 2009, Gilliam and Pleim 2010, Otte 2008 and Ngan et al 2012).

Reply: The sentence is about the studies by Ngan et al. (2012), Deng (2009), Gilliam and Pleim (2010), which only have statistics for meteorological variables, none for chemical variables. Only Otte (2008) includes the chemical statistics in her paper.

It should be rewritten as:

"However, the statistics from their study cannot be used for interpreting the sensitivity of obsnudging since its base WRF case is a forecast run which used a different analysis input " – line 90-91

3. Line 91, Daum et al., 2004

Reply: Thanks. typo, corrected

4. Line 96, missing reference for Lefer and Rappengluck 2010. Olaguer et al., 2009

Reply: Thanks. missing in v1, already corrected in online ACPD version

5. Line 103, it is good to have a citation for DISCOVER-AQ.

Reply: we added a link to the DISCOVER-AQ website (from which all the data were collected)

6. Line 141-143, suggesting not to use "OA" to refer observational nudging. Instead, just use "no-FDDA" .vs. "FDDA"

Reply: Because the base case performed the "standard" grid nudging, so it did use "FDDA". Hence it seems "no-FDDA" is not proper for base case. Please see our revisions outlined in item 1 reply.

7. Line 171, give citation for those prior modeling studies.

Reply: added citations.

8. Line 202-224, these three paragraphs should be shortened since this is not project report and technical note. Please summarize what are the data frequency for analysis nudging, surface nudging and observational nudging, what variables to be nudged, and for what vertical layers.

Reply: Thanks for the suggestions. Modified (shortened) as suggested.

9. Line 242, missing reference for Willmott 1981

Reply: Thanks. Added.

10. Line 306, it is more desirable to see time-series of hourly temperature and ozone instead of daily average since both variables have strong diurnal variation. What did the authors choose to show the daily average plots?

Reply: In the past we usually use daily average plots to check some important meteorological and chemical features during the simulated period, but we agree with reviewer's point and replaced the daily time series with hourly time series.

As a result, all the texts related to the two plots are modified. Updated figure 3 and 4 are shown below.



11. Line 378, Figure 7 is hardly to read due to poor figure quality and small text.

Reply: Thanks. Already changed in ACPD online version, should be OK now.

12. Line 394, it should be showing hourly ozone plot instead of daily average. There are a lot variations for ozone through a day.

Reply: Thanks. We replaced the plot as the reviewer suggested.

13. Line 484, Did the nudged met. data provide better ozone results than the base case in the comparison with aircraft measurements?

Reply: Yes, the nudge case is better. In the two plots (Fig 11 and 12), we showed spatial ozone for base case as background. Fig.11 is intended to show the high ozone aloft in early morning, which contributed to later model's underprediction. The comparison of model vs aircraft ozone is given in Figure 13 and 14.

14. Line 487, missing reference for Li and Rappengluck 2014

Reply: Thanks. added in ACPD online version

15. Line 500, the model under-predicted ozone both at the surface and aloft on Sep 25th. Even with nudged meteorology, there was not much improvement. Is that because the met data are still wrong or emission data may have problem?

Reply: We think that the missed high ozone on Sep 25th is a combination of transport (high observed ozone aloft in early morning), not-so-good meteorology even after nudging (even though we show some improvements in this study), and possibly unreported emission upsets.

We already identified a problem in current WRF OA process and developed new processes to correct the problem and results are surprisingly good. In the paper, the ongoing study is briefly mentioned in the last section.

16. Line 543-548, the discussion about the impact of nudging on cloud/precipitation prediction is ambiguous. There is no comparison on these two variables shown. Did the nudging configuration help to prevent the inaccuracy of the prediction or make it worse?

Reply: We felt compelled to mention cloud/precipitation since they heavily impact ozone and performing obs-nudging certainly altered the two variables (despite that nudging coefficient is set to zero for moisture). Yet we do not have good observations to quantitatively analyze the changes brought by obs-nudging. It entails another study to analyze the impact of nudging on cloud/precipitation and how ozone might be ultimately affected. 17. Line 555, what does this "small-scale meteorological events" refer to? In what sense it is relevant to the high ozone events? Is this something for future works? The conclusion section is not clear. Suggest to revise and include future works.

Reply: Thanks for your suggestions. Small-scale events are discussed briefly in first paragraph of page 5(27361) of ACPD online version, with a few references. The conclusion section has been substantially modified in the online version. We value the comment and added future works.

The last section of Discussion now reads:

"Small-scale meteorological events are frequently cited for their contributions to high ozone events. Model's capability in reproducing these events is critical in simulating such high ozone episodes. The base case did not recreate the 25 September small-scale events likely due to the complex winds and a lack of local information which can be used to steer model state closer to reality. On the other hand, the inability of the sensitivity case to replicate the local winds is likely a result of the imperfection of the nudging process pending further investigation. An ongoing study by the current authors suggests that errors in the metrological fields from the default grid nudging files are important sources. Methods are being tested to improve the quality of grid nudging files. Early results showed that the bay breeze which caused the wind reversal around La Porte was well captured through improved grid nudging files. In addition, more observational data (e.g., more sites and higher data frequency) and more testing on the combination of nudging setting should help improve the obs-nudging performance. Also, the impact of obs-nudging on precipitation and clouds should be further investigated to understand their chain effect on chemistry." **Reviewer 2**

Original comments by reviewer 2:

Specifically, the sentences are ambiguous, incomplete, and awkward throughout the text. One has to go over several times for many sentences to guess what the authors are trying to say. The amount of corrections needed is beyond what a reviewer can suggest in details. I am listening a few typical issues below.

It should have been clearly stated early on what exactly the study is trying to accomplish; what variables they "nudged" exactly, and what results they examined. A reader should not have to read through all the details of the model setups to find out what variables they actually "nudged". Statements such as ". . . the impact of OA on the simulated meteorology and ozone concentrations. . ." or ". . . indicated that OA improved the timing of wind transition . . .", are throughout the paper without indicating OA on what, or nudging what.

Another issue is that this manuscript was not written for more general readers, terminologies were used without providing background. They never explicitly explain the connection between WRF and CMAQ before using WRF-CMAQ. The terms nudging and OA were used interchangeably without explaining the differences.

General response:

We thank the reviewer for his/her input. As for scientific significance, we think the paper has several findings not seen before in the previous studies (regarding more detailed impact of objective analysis on meteorology including temperature, winds and PBL height and chemical concentrations of ozone). And these findings are important for later works in their effort to improve the WRF's nudging process. In the revised manuscript, we worked to clearly explain about how the objective analysis improved the performance of WRF and WRF-CMAQ simulations. For example, we showed how the nudging process improved the meteorology and chemical concentrations on the 25 September (Our group worked as air quality forecasting group for DISCOVER-AQ project performed in Houston in September of 2013. As we authors acknowledged, none of the previous forecasting/modeling exercises from a few modeling groups couldn't make a reasonable simulation of ozone on September 25).

In this study, we showed that objective analysis approach significantly improved meteorology for September and it also improved chemical concentrations of ozone, but the order being improved is smaller than that of meteorology. Further, we discussed what would be another cause for the uncertainty in the ozone simulation.

On the language issue, we acknowledge the paper was heavy and occasionally hard to follow. As the reviewer suggested, we, authors did proofread all the text again and rewrote/modified the significant amounts of the contents from the beginning to the end and we believe now that the English issue the second reviewer suggested was resolved. For example, over a hundred changes have been made from abstract to conclusion of the revised manuscript.

Itemized response:

Since the reviewer's comments are not with line numbers, we try to respond as best as we can.

- 1. One of the major concerns is the contribution of the paper to science. Here we elaborate a few points.
 - a. Meteorology is the foundation for emission and chemistry studies. Without a good set meteorology when studying real-world air quality, one can hardly draw conclusions with. As we addressed above, a good example is the 09/25 episode discussed in the paper, we still cannot say whether an emission event or unknown large-scale transport played major roles because meteorology is not well simulated. It was our motivation for this study in the beginning.
 - b. FDDA is a critical tool in improving model performance in meteorology modeling. A large portion of performance gain in the last 30 years came from FDDA. Objective analysis (OA) and observation nudging are critical methods in FDDA.
 - c. We have not found any paper providing a quantitative sensitivity study on the possible performance gain from OA and observation nudging on both meteorology and chemistry. Ngan et al. (2012) is the closest in

it evaluating both meteorology and chemistry yet it is not a sensitivity study on obs-nudging.

- d. We have not found any other air quality study performing 1-hr observation nudging.
- e. More importantly, the paper provided an example showing the failure of model to replicate the high ozone even after OA and observation nudging are performed. Implicitly, this means that trying different physical schemes is unlikely to solve the issue (we did test a few cases although this statement largely came from our years experience in air quality modeling). This begs for the question: what can we do to recreate the right meteorology so that we can produce high ozone in chemistry model? As we addressed in the discussion of the revised version, we have been working to develop a novel technology (a new nudging process) to address the issue (preliminary results are surprisingly encouraging).
- f. We evaluated model PBL and ozone aloft to study the model sensitivity to obs-nudging and OA which we have not seen before.
- 2. About nudging technique and related terminology.
 - a. The authors expected the readers to have basic knowledge on FDDA and nudging, i.e., what is nudging, what it for is, etc. We didn't provide further explanations on certain details due to the large size of the revised manuscript. However, we added some explanation in the revised version, e.g., FDDA in line 49-50, Objective Analysis in line 70-72.
 - b. We have updated the manuscript to clarify the terminology, especially what we did in addition to standard grid nudging. We assimilated observations from various sources into the simulation. WRF has a separate program "OBSGRID" to perform objective analysis (OA) and create input for observation nudging. Therefore, performing OA generally means doing observation nudging simultaneously to make the most use of observations. However, the term "OA and observation nudging" is too cumbersome. Therefore, in revised manuscript, we used "observation nudging" and explained that it means the combined "OA and observation nudging" – line 72-75.

The changes on obs-nudging and OA are reflected throughout the text.

- c. OA is to improve first guess meteorology analysis by incorporating observations we included this statement in the revised version, line 70-72.
- d. WRF observation nudging is performed on 4 variables, U, V, T, Q (water vapor mixing ratio), line 228-229. The default is nudging all. For grid nudging, there are two extra variables (geopotential and dry column mass) inside a grid nudging file, although there is no explicit control on the two variables in the namelist. There are concerns with nudging T and Q in some occasions that they may be turned off.

In original manuscript, we already stated we did not perform grid nudging on mass fields (T and Q) inside PBL, line 199-200. Also there is no observation nudging for "Q", line 229-231.

- 3. WRF-CMAQ and OA/observation nudging
 - a. We changed "WRF-CMAQ" to "WRF and CMAQ", line 21.
 - b. We explained the difference between OA and observation nudging. OA is to improve first guess meteorology analysis by assimilating observations, which is to improve the quality of analysis nudging. Observation nudging is to nudge model state toward observations at measurement locations – line 65-72.
- 4. The language issue

We have done a major overhaul on the language part to improve the readability, with input from several co-authors. As one can see from the track-change version, many changes were made. This is in addition to the modification suggested by the handling editor, reflected in the ACPD version.

We also made better plots to assist readability.

We agree that the paper is not a light read and it has much information, especially on the 25 September, the high ozone day. We think most of them are quite relevant but did trim some contents, such as the ozone description on the 26th.

1	The impact of observation nudging on simulated meteorology and
2	ozone concentrations during DISCOVER-AQ 2013 Texas campaign
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4 5	Xiangshang Li ¹ , Yunsoo Choi ¹ , Beata Czader ¹ , <u>Anirban Roy¹</u> , Hyuncheol Kim ^{2,3} , Barry Lefer ¹ , Shuai Pan ¹
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13	Abstract
14	Air quality modeling demands accurate Accurate meteorological simulations fields are imperative
15	for correct chemical transport modeling. Observation nudging, also known asalong with
16	objective analysis $(OA)_{r_{a}}$ is generally considered <u>as</u> a low-cost and effective technique to
17	improve meteorological simulations. However, the meteorological impact of OAobservation
18	nudging on chemistry has not been well characterized. This study involved -two simulations
19	(with/without OA) to analyze the impact of OA observation nudging on the simulated
20	meteorology and ozone concentrations during the 2013 Deriving Information on Surface
21	conditions from Column and Vertically Resolved Observations Relevant to Air Quality
22	(DISCOVER-AQ) Texas campaign period-in September 2013, using Weather Research and
23	Forecasting (WRF) and Community Multiscale Air Quality (CMAQ) models. The results
24	showed improved correlations between observed and simulated parameters from the
25	OAsensitivity case. The index of agreement (IOA) improved by about 9% for surface
26	temperature and 6-11% for surface zonal (U-WIND) and meridional (V-WIND) winds when

27	OAobservation nudging was employed. Analysis of a cold front event indicated that OAit
28	improved the timing of wind transition during the front passage. Employing OAobservation
29	nudging also reduced the model biases infor the planetary boundary height predictions. For
30	CMAQ simulated simulations, the IOA improved by 6% in the sensitivity case for surface ozone
31	during the whole simulatedentire simulation period, IOA improved by 6% in the OA case. The
32	high ozone episode on September 25 th was a typical-post-front ozone event in Houston. The
33	small-scale morning wind-shifts near the Houston Ship Channel combined with higher aloft
34	ozone from recirculationearly morning likely caused the day's ozone exceedance. While
35	OAobservation nudging did not reproduce recreate the wind shifts on that day and failed to
36	reproduce the observed surface and aloft high ozone, analyses of surface and aircraft data found
37	that OA results matched better with observationsobservation nudging helped model to yield
38	improved ozone predictions. In a two-hour period during the event, substantially better winds in
39	OAthe sensitivity case noticeably improved the ozone. Further work on improving OA'sthe
40	capability of nudging to reproduce local meteorological events could enhance a
41	chemistrychemical transport model's abilitycapability to predict high ozone events.
42	Keywords: WRF, CMAQ, air quality model, DISCOVER-AQ, observation nudging

44 **1. Introduction**

45 Accurate meteorological simulations are essential to photochemical modeling since meteorological Meteorological variables, such as cloud fraction, winds, planetary boundary layer 46 47 (PBL) heights and precipitation; significantly impact air quality. They influence the production, 48 transport, and deposition of various chemical species (e.g., Pour-Biazar et al. 2007; Banta et al. 49 2005; Cuchiara et al. 2014). Hence accurate meteorological inputs are imperative for air quality 50 modeling. Common approaches of improving meteorological simulations include the selection of 51 updated and high qualityresolution terrain and input data (e.g., Cheng and Byun 2008), the 52 optimization of physics and dynamics options (e.g., Zhong et al. $2007_{\frac{1}{2}}$) and the implementation 53 of four dimensional data assimilation (FDDA). The air quality modeling group at the University 54 of Houston (UH) had performed several sensitivity studies on the various parameterization 55 schemes in the recent past (e.g. Zhong et al. 2007; Ngan et al. 2012; Cuchiara et al. 2014).

56	FDDA continuously merges new observational data into model simulation such that the model's
57	predictions do not drift away from observations. There are several FDDA methods including
58	nudging (e.g., Stauffer and Seaman 2004) and Variational Methods (3D-VAR or 4D-VAR) (:
59	e.g., Le Dimet and Talagrand 1986; Huang et al. 2009). 4D-VAR obtains optimal states of the
60	atmosphere using multi-time-level observations by globally adjusting a model solution to all
61	available observations over an interval of time. Nudging is a simple yet flexible FDDA method
62	originally developed by Stauffer and Seaman (1990, 1994), and implemented in the Fifth-
63	Generation PSU/NCAR Mesoscale Model (MM5). Not intended for optimal adjustment, nudging
64	is less computationally intensive andbut needs special care for the nudging coefficients. Nudging
65	involves adding an artificial tendency term to one or more model prognostic equations that
66	reflect the difference between the best estimate of the observed state and the model state at a
67	given location and time. In short, the goal is to "nudge" model state towards observed state.
68	There are several types of nudging such as 3D analysis nudging, surface analysis nudging, and
69	observation nudging. (obs-nudging). In the case of analysis nudging, the model state is nudged
70	toward gridded analysis. The difference between 3D and surface analysis nudging is that 3D
71	analysis (at all model levels except for surface) data are used to improve 3D fields while surface
72	analysis data are used to improve surface fields. In observation nudging, the model is perturbed
73	such that its predictions are nudged to match better with observations at individual locations,
74	both on-surface and aloft. The MM5 nudging codes were later improved and incorporated into
75	the Weather Research and Forecasting (WRF) model by Liu et al. (2005, 2006). The
76	enhancements enable observation nudging to assimilate a large variety of direct or derived
77	observations. obs-nudging to assimilate a large variety of direct or derived observations. In
78	WRF, the inputs for obs-nudging are generated by WRF OBSGRID program. This program also
79	performs Objective Analysis (OA) to improve the quality of analysis nudging files. Objective
80	Analysis updates first guess meteorology analysis by incorporating observational data. Since
81	obs-nudging is usually performed along with OA (as in this study) to maximize the benefits of
82	assimilating observations, we also use OA to denote the combined Objective Analysis and obs-
83	nudging processes in case names.
84	The benefit of applying nudging to improve meteorological simulations has been demonstrated

The benefit of applying nudging to improve meteorological simulations has been demonstrated in many studies (e.g., Deng 2009; Gilliam and Pleim 2010). However, only a few have extended the investigation into chemistry impact of the improved fields on air quality simulations- has been

87	investigated by relatively fewer studies. Otte (2008) showed that analysis nudging is able to
88	improve MM5 meteorology, as well asthe Community Multiscale Air Quality (CMAQ) model
89	with improved MM5 meteorology using analysis nudging was able to better simulate ozone
90	chemistry as reflected in ozone-model-measurement statistics. Better Their results indicated that
91	better "model skill" scores were achieved for daily maximum 1-hr ozone mixing rationations after
92	analysis nudging over a 35-day periodsimulation episode. Byun et al. (2008) performed over a
93	dozen tests on observation obs-nudging (with analysis nudging turned on) and showed
94	observation obs-nudging improved both winds and temperature in MM5 simulations. The study
95	also gave an example in which improved wind fields on a given day ledhelped the CMAQ
96	tomodel better capture the high-ozone areahotspot southwest of Houston. Ngan et al. (2012)
97	compared results from several MM5-CMAQ simulations and showed coupled to the MM5 model
98	which included nudging. Their results indicated that fully nudged (with both analysis nudging
99	and observation obs-nudging implemented) simulations outperformed a forecast run in
100	performed better with respect to both meteorology and ozone chemistry. Their study location
101	was Houston Texas, the same as in this study. No detailed However, the statistics were presented
102	on the quantitative improvements from the nudgingtheir study cannot be used for interpreting the
103	sensitivity of obs-nudging since its base WRF case is a forecast run which used a different
104	analysis input. Previous workstudies by the current authors (e.g., Rappenglueck et al. 2011;
105	Czader et al. 2013) showed that observation obs-nudging helps to helped correct errors in model
106	wind fields, which are c ritical to the transport process of air pollutants , as well as the and
107	production of secondary pollutants. To the best of the authors' knowledge, there is no
108	comprehensive existing study on the impact of observation
109	whenair quality simulation using the meteorologicalWRF model is WRF
110	This study intends to fill up the gap in the studies mentioned above by investigating the
111	sensitivity of WRF- <u>and subsequently</u> , CMAQ simulations to the use of observation nudging.
112	Although not elaborated here, the WRF CMAO sensitivity to different observation nudging
112	frequencies was also explored. In theory, higher frequency of observation obs-nudging input
114	should have a higher probability to capture small scale events, such as local wind shifts. These
114	events may only slightly impact local weather, yet they can have a largemarked effect on
116	chemistry since it. This is well known that because local stagnation and wind
110	chemistry since it <u>rine</u> is well known that <u>eeeuuse</u> local staghation and wind

117 convergence/reversals can contribute to the pollutant build-up-(e.g., as indicated by Banta et al. 118 (1998;), Cheung and Wang (2001;) and Tucker et al. (2010). 119 There is a significant presence of petro chemical petrochemical facilities, power plants and motor 120 vehicles in the Houston-Galveston-Brazoria (HGB) region located in southeastern Texas 121 (SETX). The major pollutant in the region is ozone due to the abundant emissions of precursors 122 like nitrogen oxide (NO_x) and Volatile Organic Compounds (VOCs). During the long and hot 123 summer, ozone concentrations often rises rise above the threshold level as stipulated in the 124 National Ambient Air Quality StandardsStandard (NAAQS). Consequently, HGB has been 125 designated as an ozone non attainmentnonattainment region by the US Environmental Protection Agency (USEPA); (http://www3.epa.gov/airquality/greenbook/hncs.html#TEXAS). The petro-126 127 ehemical plants are largely concentrated in the Houston Ship Channel (HSC) area -128 just north of the Galveston Bay. The VOCs emitted from the HSC area are highly reactive and have been shown to contribute greatly to the high regional ozone episodes in HGB-(e.g. 129 130 Kleinman et al. 2002; Daum et al. 2003). Depending on the local meteorology, the plumes from 131 HSC may be carried to different locations in HGB and trigger high ozone events on its path. 132 Metropolitan Houston has a high level of NO_x emissions partly due to heavy vehicular traffic in 133 the city. As a result of the large amount of precursor emissions and favorable weather, relatively 134 frequent high ozone events occur in the area. Ngan and Byun (2011) gave an analysis on the 135 relationships between the high ozone frequency and underlying weather patterns. They derived the weather patterns from a classification scheme using large scale 850 hPa synoptic flow as 136 137 input. 138 TheDue to the reasons listed above, the Houston-Galveston-Brazoria region has been the 139 location focus of interest of manyseveral air quality studies in the recent past (e.g., Banta et al. 140 2005; Parrish et al. 2009; Lefer and Rappengluck 2010; Olaguer et al. 2013; Czader et al. 2013, 141 Choi et al. 2012; Choi 2014; Choi and Souri, 2015; Pan et al. 2015). It is a good place for 142 studying ozone production and transport due to the existence of a dense surface monitoring 143 network, as well as several intensive measurement field campaigns which provide ample 144 observational data. For example, in September 2013, the National Aeronautics and Space 145 Administration (NASA), joined by a number of agencies and universities, conducted a field 146 measurement campaign in SETX as part of its the Deriving Information on Surface conditions

147	from Column and Vertically Resolved Observations Relevant to Air Quality (DISCOVER-AQ)
148	program. The NASA (http://www-air.larc.nasa.gov/missions/discover-aq/discover-aq.html).
149	This program has conducted several air quality and meteorology measurements at several
150	different locations in the U.S. The availability of dense surface observations is important in OA's
151	eapability for obs-nudging to correct erroneous local winds in the model. Without a rich set of
152	observations, the The performance of OAobs-nudging will be handicapped without a rich set of
153	observations.
154	This study involved performing two sets of WRF- and CMAQ model simulations for the 2013
155	DISCOVER-AQ Texas time period in order to understand the impact of observation-obs-nudging
156	using comprehensive. The data for nudging included multiple sets of observation data from both
157	in-situ-surface and aircraftaloft measurements. We evaluated model-measurement performance
158	and calculated statistics for both WRF and CMAQ. Meteorological fields critical to ozone
159	chemistry were examined to explore the model sensitivity to OA. The paper is structured as
160	following: Section 1 is introduction; Section 2 describes the measurement data and the modeling
161	system; Section 3 covers the evaluation protocols; Section 4 discusses the general meteorological
162	conditions that occurred during the campaign period; Section 5 presents the modeling results,
163	and Section 6 provides discussions and conclusions. output.
164	
165	2. Observational Data and Model Configurations
166	For evaluation of the results, this study used regular measurements from the Continuous Ambient
167	Monitoring Station (CAMS), operated by the Texas Commission on Environmental Quality
168	(TCEQ), as well as PBL and aloft ozone measurements from DISCOVER AQ campaign. For
169	observation nudging, in addition to the CAMS data sets, several datastreams from the
170	Meteorological Assimilation Data Ingest System (MADIS) were also used.
171	2.1 Observational Data
172	This study used regular measurements from the Continuous Ambient Monitoring Stations
173	(CAMS) operated by the Texas Commission on Environmental Quality (TCEQ). Additionally,
174	PBL and aloft ozone measurement data were obtained from the DISCOVER-AQ campaign. For

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	175	observation nudging,	CAMS data and severa	al data streams from the Meteoro	logical Assimilation
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- 176 Data Ingest System (MADIS) were used. The CAMS measurement network collected real-time
- 177 meteorology and chemistrypollutant data. The measured parameters differ from station to station.
- 178 The station density at <u>South East Texas (SETX)</u> is relatively high. There are were 63 sites
- 179 having with meteorological measurements and 52 sites having with ozone measurements in the 4-
- 180 km domain-(<u>, The network is represented in Figure 1) during DISCOVER AQ time period</u>. The
- 181 stationssites are represented by dots, with the La Porte (C556) site labeled. All CAMS
- 182 observations are accessible at TCEQ website: http://www.teeq.state.tx.us/egi-
- 183 bin/compliance/monops/daily_summary.pl.http://www.tceq.state.tx.us/cgi-
- 184 <u>bin/compliance/monops/daily_summary.pl.</u>



- 185 186
- 187 **Figure 1**. Locations of CAMS sites (dots) in CMAQ 4-km modeling domain during September
- 188 2013. Metro Houston, Houston Ship Channel, Galveston Bay and Gulf of Mexico are labeled.
- 189 Additionally, PBL height measurements for September were obtained from a teamrecorded at a
- 190 site at the University of Houston, which employed LIDAR (. The PBL height was measured
- 191 <u>using the Light Detection and Ranging) to detect the PBL height. Presently, only (LIDAR)</u>
- 192 system. The PBL data at one site is currently available.

193	<u>only at this site.</u> For analysis of ozone aloft on September 25 ozone, we also used measurements
	,

- 194 from aircraft P-3B, part of the rich datasets collected during DISCOVER-AQ campaign
- 195 (http://www_air.larc.nasa.gov/missions/discover_aq/discover_aq.html)._ The P-3B data had over
- 196 100 parameters and which are accessible from the website online.

197 2.2 Model Configurations

- 198 The modeling system consists of <u>the WRF-meteorological model (Skamarock et al., 2008), the</u>
- 199 Sparse Matrix Operator Kernel Emissions (SMOKE-) model for emissions modeling (Houyoux
- 200 et al., 2000) and the CMAQ models as described model (Byun and Schere, 2006) for chemical
- 201 <u>transport modeling. The details about model configurations are presented</u> in the following three
- 202 subsections sections. Two sets of simulations, were conducted, one set with the only difference in
- 203 whetherobs-nudging and OA was adopted, were performed and the other without. The base case,
- 204 referred as "No-OA", did not employ observation obs-nudging or OA. The second case, "1Hr-
- 205 OA", performed observation-obs-nudging and OA using hourly observationnudging input.
- 206

2.2.1. WRF ConfigurationsSetup

Both WRF simulations used the same nested domain and NARR (North American RegionalReanalysis) as input, with grid nudging turned on.

209 2.2.1.1. Domain Setup

210	Figure 2 depicts the horizontal domain setup. There were two <u>Two</u> nested domains were used,
211	with 12-km and 4-km resolution respectively. The 4-km domain covered SETX and a small
212	portion of Louisiana. The 12-km domain (red box) encompassed Texas and parts of a few
213	neighboring states (or parts) The number of grid sizescells for the 12-km and 4-km domains
214	were 161×145 (E-W by N-S), and 95×77 respectively. The projection type iswas Lambert conic
215	conformal (LCC). Three projection parameters, were considered: namely first latitude, the
216	(33°N), second latitude (45°N) and the standard longitude, are 33°N, 45°N and (97°W degrees)
217	respectively.). The USEPA used the same projection parameters to develop emission inventories
218	for air quality modeling. Vertically both (Mason et al. 2010). Both domains had a vertical
219	resolution of 27 eta layers based on dry hydrostatic pressures. The model top is set to be 100 hPa
220	corresponding to top layer pressure of the input NARR data.

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TRIN<u>T RIII</u> WIRI and CIMAQ).

226

2.2.1.<u>21</u>. Input Data

227	BothThe NARR data used for WRF simulations were retrospective runs using NARR analysis as
228	input, <u>is</u> downloadable from: http://rda.ucar.edu/datasets/ds608.0/. The NARR
229	http://rda.ucar.edu/datasets/ds608.0/. The data were based on an Eta 221 grid at 29 pressure
230	levels. Its horizontal resolution was 32-km and the frequency was 3hourly. The initial and
231	boundary conditions were generated from the NARR analysis by WRF. An alternative to NARR
232	was the Eta-NAM analysis data. However, the data temporal frequency was lowered from 3-
233	hourly to 6-hourly starting 2013. Our testtests showed that it was not as good as NARR for WRF
234	input, dataset - likely because of lower temporal resolution.

235 2.2.1.32. Physics and FDDA Options

236 Major physics options were-used in the model are listed in Table 1. These options are consistent 237 with the WRF options in our daily air quality forecasting system (http://spock.geosc.uh.edu/). 238 Among them, the PBL and cumulous cloud schemes are especially critical. Our past modeling 239 experiences demonstrated indicated that employing the Yonsei University (YSU) is the bestfor PBL scheme in Houston case study while and the Kain-Fritsch (K-F) is the preferable for 240 241 cumulous scheme. The choice of gave the best results for the Houston area. YSU scheme iswas 242 also corroborated recentlyone of the two PBL schemes recommended by Cuchiara et al. (2014) 243 et al...). The K-F scheme is "drier" than others and produces less bogusnumber of "false" 244 convectional thunderstorms. The numbers in parentheses represent the value of corresponding 245 namelist variable in WRF's namelist file. For example, the "1" after YSU is the value of the 246 namelist variable "bl pbl physics" in WRF's namelist file. For both of the simulations, we performed standard grid nudging for both of the cases using NARR analysis. For grid nudging 247 options, we generally followed the recommendations in the WRF's User Guide. For example, the 248 249 mass fields (temperature and moisture) were nudged only at layers above the PBL while wind 250 fields were adjusted at all layers including the surface layer.

251 2.2.1.4<u>3</u>. Observation Nudging with MADIS and CAMS data in WRF

As mentioned in the introduction, observation nudging is regarded as a low-cost and effective
 method for improving meteorological model performance, but it requires additional<u>Additional</u>
 observational data. In this study, we acquired are required to implement obs-nudging and OA. To
 generate the input files for the OBSGRID program, we processed the observation data and

- 256 generating files in "little_r" format using similar procedures found in the approach of Ngan et al.
- 257 (2012) and Czader et al. (2013). Observational data came from the MADIS and TCEQ CAMS.
- 258 MADIS (https://madis.ncep.noaa.gov/); is a National Oceanic and Atmospheric Administration
- 259 (NOAA) program, which collects, integrates, quality-controls, and distributes observations from

260 NOAA and other organizations. <u>Additional information is available online</u>,

- 261 <u>https://madis.ncep.noaa.gov/.</u> The four MADIS datasets used for observation-obs-nudging were
- 262 NOAA Profiler Network (NPN), Cooperative Agency Profilers (CAP), Meteorological Terminal
- 263 Aviation Routine (METAR) weather report and NOAA Radiosonde (RAOB). The METAR
- 264 dataset was collected by mostly first-order, METAR reporting, surface monitoring stations.
- 265 NPN, RAOB and CAP were the most commonly used upper air datasets.

266 The "little_r" files from previous stepprocessed input observation data were fed into WRF

- 267 OBSGRID module to update the domain analyses ("met em" files), and, generate additional
- 268 surface analyses ("sffdda") and text nudging files ("OBS_DOMAIN"). Actual observation obs-
- 269 nudging was performed by the main WRF program by properly setting observation after obs-
- 270 nudging namelist variable variables are properly set. The namelist for OBSGRID and relevant
- 271 WRF section settings came largely from recommended values of WRF User's Guide and a
- 272 previous study by Ngan et al. (2012).
- 273 Theoretically, observation-obs-nudging updating at a higher frequency should enhance the 274 model's performance. A typical frequency of input analysis data is 3-hourly while the frequency 275 for observational data is hourly. The 3-hourly frequency of input analyses may be the reason for 276 the default 3-hour time- interval in WRF's OBSGRID settings for generating the observation 277 obs-nudging files. Since there were few existing OAobs-nudging studies related to air quality 278 and we are not aware of any reference to the adoption of 1-hour input frequency, we assume that 279 all the existing studies used the default 3-hour interval. As the WRF model allows the interval to 280 be set to 1-hour or smaller when corresponding observational data were available, we tested both 281 1-hour and 3-hour scenarios. The results indicated that 1-hour OAobs-nudging had slightly better 282 performance than the 3-hour one. As a result, this study adopted 1-hour temporal frequency for 283 observation nudging. The quantities that were nudged were temperature, moisture, and the two 284 wind components (U-WIND and V-WIND). Obs-nudging for moisture was not performed in

285	this study. This was based on our past experiences since performing moisture nudging sometimes
286	trigger excessive artificial thunderstorms which disrupted model flow fields.
287	It should be noted that the default time interval for modified gridded analyses, i.e., the "metoa-
288	em" and "sgfdda" files have to match input analysis data in OBSGRID. The namelist variable
289	was called "interval", with a default value of "10800" seconds. The time interval for output
290	nudging files was set by namelist variable "int4d", with the same default value of "10800"
291	seconds. To output the observation nudging files hourly, "int4d" should be set to "3600"
292	seconds. This means that the OBSGRID output files, "metoa_em" and "OBS_DOMAIN", did
293	not have the same interval in our study.
294	In WRF, there were a few namelist variables controlling the frequency of grid nudging and
295	observation nudging. The first one was "interval_seconds", which should match the interval of
296	input grid nudging files ("met em"). The second one was "sgfdda_interval_m", matching the
297	interval of surface grid nudging files ("sgfdda"). In our simulation, both intervals were equal to
298	3-hours. The third one was "auxinput11_interval", controlling the updating interval for
299	observation nudging files ("OBS_DOMAIN"). The last one, "obs ionf", determined the nudging
300	frequency relative to internal integration time step. For example, if the integration time step for
301	the coarse domain is 30 seconds, setting "obs_ionf" to 1 means performing OA every 30
302	seconds, while setting "obs_ionf" to 3 means performing OA every 90 seconds. In our
303	simulation, "obs_ionf" is set to 1.
304	One departure from the default OA setting in WRF was that the moisture OA was turned off with
305	"obs_nudge_mois" set to 0. This was based on our past experiences since performing moisture
306	OA sometimes trigger excessive artificial thunderstorms which disrupted model flow fields.
307	2.2.2. EmissionEmissions Processing
308	For anthropogenic sources we utilized the National EmissionEmissions Inventory of 2008
309	(NEI2008) generated by the USEPA. The mobile (USEPA, 2011). Motor vehicle emissions for
310	this inventory were processed with EPA's using the EPA's Motor Vehicle EmissionEmissions
311	Simulator (MOVES). Using) (USEPA, 2015). The inventory was processed using the Sparse
312	Matrix Operator Kernel Emissions (SMOKE) Modeling System model v3.1 the inventory was
313	converted toto obtain gridded emission rates as well as to emission species as listed in and

314	speciated for the Carbon Bond 05 (CB05) chemical mechanism that is used for use in the CMAQ
315	modelingmodel. The biogenic emissions were estimated modeled using the Biogenic Emissions
316	Inventory System (BEIS) $\frac{1}{2}$.14. Although NEI2008 might have overestimated NO _x emissions
317	in Houston (e.g., Choi 2012; Czader et al. 2015) which could have impacted on ozone formation
318	in the region, 2015), we used base NEI2008 without adjustment because the adjustment of the
319	NO _x emission also has large uncertainty. Pan et al. (2015) showed that the CMAQ ozone
320	performance using NEI2008 appears reasonable.

321 2.2.3. CMAQ Configurations

322 The USEPA's CMAQ (Byun and Schere 2006) version 5.0.1 was adopted for this study, 323 following the choice of several other Houston. Several air quality modeling studies focusing on 324 the Houston area have used this model (e.g., Foley et al. 2010; Czader et al. 2013, 2015; Choi 325 2014; Pan et al. 2015). CMAQ horizontal domains were slightly smaller than the WRF 326 counterpart in order to avoid the discontinuity near the domain boundary. The domains were 327 shown in Figure 2 as green and brown boxes. The chemical boundary conditions for all the 328 species in the 4-km domain were derived from 12-km domain air quality forecasting results 329 (http://spock.geosc.uh.edu). Vertically, CMAQ inherited(http://spock.geosc.uh.edu). The model 330 used the same layers from vertical structure as WRF-without layer collapsing. Major CMAQ 331 configurations were described are listed in Table 2. The texts in the parentheses were the values 332 in the CMAQ build script.

Chemical processes were simulated with the <u>available in CMAQ CB05</u> chemical mechanism with <u>cloud/aqueous chemistry</u>, active chlorine chemistry and updated toluene mechanism. For aerosol modeling, the fifth-generation CMAQ aerosol mechanism (AE5) <u>withwhich includes</u> sea salt <u>ismodeling was</u> selected. <u>Cloud/aqueous chemistry is included</u>. The total number of included species is 132, with 70 reactive gas-phase, 49 aerosol and 13 non-reactive species.

338

339 3. Evaluation Metrics

To assess model performance against observations, we computed a set of five statistics including
Pearson correlation, index of agreement (IOA, Willmott 1981), mean bias (MB), root mean

342	square error (RMSE), and Mean Absolute Error (MAE),). This list is similar to one used by Li et	
343	al (2008)-) for model performance evaluation. The goal is to have a comprehensive comparison	
344	between model and observation time series. These The set of five statistics have been frequently	
345	used for performance evaluation in modeling community.was divided into three groups:	
346	The set1) Evaluation of five statistics was divided into three groups:	
347	1) Measuring the direct departuremagnitude of model results from observation, vis-a-vis in	
348	measurement units-situ data	
349	• Mean Bias (MB)	
350	Mean Absolute Error (MAE)	
351	Root Mean Square Error (RMSE)	
352	2) Measuring how close the model values follow changes in the observations, unitless	
353	Correlation	
354	3) A composite performance index, index of agreement (IOA or d) suggested by Willmott	
355	(1981), unitless	
356	IOA is considered a better performance index than correlation as it takes into account the	
357	difference in the means and standard deviation. For example, when correlations are similar,	
358	lower model biases would yield higher IOA values. Additionally, the mean and the standard	
359	deviation of model values and observations were included as a reference.	
360	Additionally, the mean and the standard deviation (Std. Dev.) of model values and observations	
361	were included as a reference.	
362		
363	4. General Meteorological and Ozone Conditions in September 2013	
364	The weather during the-September 2013 simulation period-was relatively dry with mostly	
365	southerly, easterly or southeasterly winds. From 09/055 to 09/19 (all dates are in MM/DD	
366	format), September, there was a lack of influence of strong synoptic weather systems. Shifting	



376 precipitationwarming continued in the next few days until the 28th.





Figure 3. Regional dailyhourly temperature averaged over all available (typically around 1200)
hourly CAMS observations, two model cases also included for September of 2013.



In SETX, highHigh ozone events in SETX during fall season wereare typically associated with a
passage of cold front (e.g., Rappenglueck et al. 2008). The only ozone event with hourly surface
ozone exceeding 120 ppb (parts per billion) in September, which occurred on the 25th, fell in this
category.

393 Figure 4 showsplots the dailyhourly regional averaged ozone. On most days, the observedin-situ 394 averaged ozone fellconcentrations were below $\frac{3070}{200}$ ppb. Since the winds after dawn consistently 395 pushed the precursors from the industrial area to the southwest of the city, the wind pattern did 396 not favor the local ozone production. The daytime winds also contained a persistent easterly 397 component which moved the pollutants away from the Houston metropolitan area. In the first 10-398 day period, lesslow background ozone originating from the Gulf of Mexico contributed to the low-ozone days. With overcast skies on the 19th and the 20th, hourly high ozone values dipped 399 below 2030 ppb. The two highest ozone days; characterized by post-frontal ozone events; were 400 the 25th and the 26th. 401







405



406

407

<u>5.</u> Evaluation of Simulation Results

To evaluate the WRF simulation, we calculated statistics for surface temperature and winds in the 4-km domain. For PBL heights, we chose to plot out the time-series for the one site we had observations due to significant amount of missing data (data coverage is about 50%). For CMAQ evaluation, we calculated the surface ozone statistics for the whole month. Also, we plotted vertical ozone profile and calculated biases for <u>aloft</u> ozone <u>aloft</u> on <u>09/25.the 25th</u>.

413 **5.1.** Meteorology

414 **5.1.1. Temperature**

415	The comparison of regional averaged daily temperatures average hourly temperature for the
416	analyzed timesimulation period is shown in Figure 3. The regional observed averaged
417	dailysurface temperature was calculated by averaging the hourly temperature from ~60 CAMS
418	sites in the 4-km model domain. Despite the differences in The base case temperature was too
419	high compared to the in-situ measurements. For example, the days with more
420	clouds/precipitation,-No-OA maximum temperature for the simulated averaged temperatures
421	tracked21 st was 30°C compared to 25°C for the in-situ data-very well. It was also evident that.

Formatted: List Paragraph, Outline numbered + Level: 1 + Numbering Style: 1, 2, 3, ... + Start at: 1 + Alignment: Left + Aligned at: 0.63 cm + Indent at: 1.27 cm The high biases in the base case are sharply reduced in the "1Hr-OA" case and temperature
matched better with the observations for several time periods, especially for September 20-23.

424 __The statistics of hourly surface temperature are <u>presented_listed</u> in Table 3. With higher IOA and 425 lower mean biases (MB), the "1Hr-OA" case was clearly better than the base case "No-OA". The 426 IOA of "1Hr-OA" was about 9% higher than the base case.

427 **5.1.2. Winds**

428 Wind fields are known to significantly affect chemistry (e.g., Banta et al. 2005, 2011; Darby 429 $\frac{2005}{100}$. In ozone chemistry, winds affect the accumulation of precursors and hence the resulting 430 ozone production. Winds (e.g., Banta et al. 2005, 2011; Darby 2005). They are also responsible 431 for dispersing high ozone and bringing in background ozone. In HGB, prevailing Prevailing 432 summer time southerly to southeasterly winds in the summer timeHGB region significantly 433 loweredlower the ozone levelconcentrations in the metrometropolitan area. Therefore, high 434 ozone events usually occur when such wind pattern changespatterns change. Cold front 435 intrusion, intrusions coming as early as late August, blows blow pollutants to the south. As a result, an area of high ozone develops in the Gulf. A few days later, Following cold fronts 436 437 weakenweakening and the weather warming up, reversing winds can bring high ozone back to 438 land. High ozone may also occursoccur during intra-day-pollutant recirculation events when 439 pollutants previously blown away from industrial zone are brought back by reversing winds. The 440 high Correctly simulating these recirculation events is particularly important in predicting the 441 high ozone event caused by post-front conditions. The ozone event in the HSC area on 09/25 was 442 likely due to a combination of local recirculation caused by onset of the bay breeze and increased 443 background ozone brought in by the much larger scale southerly flow from the Gulf.transport. 444 Due to the land-water thermal contrast and the different size of the Galveston Bay and the Gulf 445 of Mexico, the western shore of the Galveston Bay often experiences a successive onset of bay 446 breeze and sea breeze in the summer. The bay breeze is typically a weaker easterly while sea 447 breeze is a stronger southeasterly. Sea breeze usually comes one to a few hours later after the bay 448 breeze. The bay breeze and the subsequent sea breeze phenomena in Houston were described by 449 Banta et al. (2005).

451	Table 3. The purpose of choosing U and V over wind speed and direction is to avoid the
452	anomalies in the wind direction statistics. For example, although wind direction of 5 and 355
453	degrees are close, the statistics suggest that they are distinctively different.
454	For both U and V components of wind, "1Hr-OA" had higher correlation and IOA than "No-
455	OA". The model performance on U and V are similar, with the correlation in a range of 0.76 to
456	0.81 for all the cases. As a reference For comparison, the performance of the OA case ("M1") in
457	Ngan et al. (2012) is very close to that in this study, with a correlation of 0.75 for U and 0.82 for
458	V. In terms of IOA, the OA case had a larger lead over the base case, ahead by 5-6% in U and
459	10-11% in V over the base case. This can be explained by the much reduced wind biases in the
460	OA case.
461	The base case had consistently stronger winds, especially the southerly component, than the
462	observation. This was reflected in the mean bias "MB", as well as the model mean "M_M".
463	Winds were reduced significantly after OA was performed. Interestingly, the high southerly bias

The statistics of zonal (U-WIND) and meridional (V-WIND) wind components are listed in

450

464 in "No-OA" turned slightly negative after OA. Winds originating from the Gulf were also

465 stronger in base case, which played a role in raising the ozone level incomparing to the

466 area.sensitivity case. Figure 5 illustrated the slowing down of southerly winds after observation

467 nudging. As a result, winds of nudging, the wind vectors matched better to the observations.





470 **Figure 5**. Model and observed winds at 09/01_00 CST: No-OA (top) and 1Hr-OA (bottom).

471 Model winds are blue arrows and the observations are orange arrows. Stronger southerly winds,

472 especially along coastal region, were reduced in the OA case.

473 **5.1.3. PBL height**

474 Atmospheric pollutants are largely confined in the PBL as most of the emissions sources are

475 | close to the ground level. <u>Hence the PBL height</u> plays a critical role in mixing and spreading the

476 pollutants. Haman et al. (2014) studied the relationship between ozone level and PBL height at a

477 Houston CAMS site and found that nighttime and early morning PBL heights were consistently

478 lower on high ozone days than on low ozone days. Czader et al. (2013) pointed out that the

479	model underprediction of PBL during nighttime may have caused the CO overprediction at the
480	same site. CO is a good proxy for understanding model's transport since it has low reactivity and
481	a relatively long life time in the troposphere.
482	_Cuchiara el al. (2014) conducted four WRF/Chem sensitivity tests on the using different PBL
483	schemes over southeast Texas. While no preferred PBL scheme was identified for WRF
484	simulations, the Yonsei University (YSU) scheme outperformed performed better than others in
485	terms of ozone prediction. As a note, we used YSU in this study as it had been tested in the past
486	and the study by Cuchiara et al. (2014).
487	The PBL height data were taken at an urban site very close to CAMS site C695, located on
488	University of Houston campus. A study by Haman et al. (2012) showed that Houston'sthe daily
489	maximum PBL height at the University of Houston site indicated previously reached its highest
490	values of slightly over 2000 m in August. In September, typical daily maximum PBL height was
491	1500 m at 15 CST while daily minimum was just below 200 m between 00 CST and 06 CST.
492	The comparison of observed and model PBL height is shown at Figure 6. The Our results
493	indicated that the model tended to overpredict the daily maximum and OAPBL height; obs-
494	nudging helped to reduce the overpredictions overprediction. For the daily minimum PBL height,
495	"No-OA" case had slightly high biases while the OA case matched quite well with
496	observationsin-situ height data. The observed minimum PBL height was lower than that reported
497	by Haman et al. (2012), likely due to the cloudy condition <u>conditions prevailing</u> in September
498	2013. There was no apparent explanation on the reduced daytime PBL biases in the OA case
499	than the base case, but it is likely the results of improved winds and temperatures in PBL.



501 Figure 6. Planetary Boundary Layer (PBL) height time series at CAMS C695 for September502 2013.

503 5.1.4. Cold Front Passage

- 504 The surface winds on 09/20 were overwhelmingly southerly in the region and reversed on $\frac{09/21}{21}$
- 505 September due to the arrival of a cold front. The hour-by-hour wind shifts for 11 sites in HGB on
- 506 09/21 <u>September</u> are plotted in Figure 7. The sites are sorted by latitude with the southernmost
- 507 site, Galveston C1034, located at the bottom row. There was only one site, Deer Park C35,
- 508 showing weak southerly at 00 CST while all the others had mostly weak northerly. Starting from
- 509 01 CST, winds in the entire HGB area turned northerly to northeasterly and continued gaining
- 510 strength in the next few hours, indicating cold air had taken over the region.
- 511 Both cases performed reasonably well on 09/21 and the timing of wind shift was captured quite
- 512 | accurately; although "No-OA" lagged about an hour.behind by ~ 1 hr. The winds turned weak
- 513 northerly at 00 CST for most sites andbut the "No-OA" case still showed the wind direction to be
- 514 all southerly. Besides the timing, OA also helped moderate the winds as the northeasterly winds
- 515 in "No-OA" case sometimes were too strong; obs-nudging helped moderate the winds. The
- 516 reduced V-wind bias in "1Hr-OA" was also evident in the wind model-measurement statistics on
- 517 09/21 is reduced from -2.5 m/s to -0.6 m/s after OA was performed. The performance of the OA

518 case during cold front passage was consistent with our past simulations. September.



520 **Figure 7**. Hourly model (blue) and CAMS (orange) winds at 11 sites on <u>09421 September</u>: No-

521 OA (top) and 1hr-OA (bottom). The 1hr-OA case is better in 00 CST to 02 CST and 17 CST to

522 20 CST.

519

523 5.2. Ozone

524 **5.2.1.** Regional Daily Average Hourly Ozone

525 Figure 4 showed plots the regional average dailyhourly ozone, which was defined similarly to 526 averaged daily the average temperature. Regional averaged daily ozone provides a global view on model's performance. Model failure of daily averaged ozone (such as wrong trend or too high 527 528 bias) was often a sign of model flaws. For example, a consistently high ozone bias could mean either the model background ozone or the emission of the precursors are too high. On the other 529 530 hand, if the high biases are present only at certain days, then it is likely a meteorological problem 531 than issues in model background or emission inventory. Overall, observed ozone level 532 wasconcentrations were low and the model did well on the daily trend although positive biases 533 were seen for some days. Although model had high biases for majority of the days, biases were consistently lower for the 534 ΘA case during two periods: 09/07 a reasonably good job on capturing the timing of intra-day 535 536 variations. However, both cases tended to 09/09 overpredict the daily highs and 09/17 to 09/21. The reduced biases were likely due to daily lows, especially in the first 8 days and between 15 537 and 21 September. An obvious departure is the 25th – both cases missed the daily high. During 538 the model high bias period, the OA case usually did better in reaching the daily low although it 539 overpredicted the high a bit more than the base case. The night time biases were reduced likely 540 541 because the lower southerly winds in the OA case since model had higher background transported 542 less ozone originated from the Gulf- to the land. In Figure 4, the first three orange circles showed the days with high model biases. The first two 543 544 eircles consisted of days with lower Our results suggested that the modeled ozone concentrations were likely higher in the Gulf than "normal" background ozone actual. However during the 2nd -545 4th and 7th-8th of September, the incoming ozone from the Gulf was markedly lower. Since the 546 547 model ozone had fixed boundary values, the model was unable to capture the daily ozone 548 variation at the boundary. The third circle consisted of days with model showed the highest biases during period of the 19th -20th likely due to overcast skies. The high model biases were 549

- 550 likely the result of problems and uncertainties in model's cloud fields and high background
- 551 ozone values. Despite the overprediction, the biases in OA case are notably lower during the

552 <u>nights of 19th and 20th.</u> A future study to upgrade the accuracy of cloud fraction using remote
553 sensing data (e.g., MODIS) should be helpful in explaining the biases.

554 There were a few days with elevated ozone due to post-front meteorology conditions. The only

exceedance happened on 09/25, which was likely caused by meteorological events in Houston

and the Galveston Bay. Averaged The overall ozone on 09/26 September was slightly higher after

southerly winds transported back the ozone from the Gulf, raising the ozone level in the entire

region. A more detailed analysis of model predictions on 09/25 and 09/26 will be presented in

559 following subsection of 5.2.3.

560 5.2.2. Performance Statistics

The ozone statistics were displayed<u>are listed</u> in Table 4. Both cases had very close correlation of 0.72 and 0.73. However, the mean biases in the OA case were lower by 3.2 ppb, which helped raise the IOA from 0.78 to 0.83. The model standard deviation increased in the OA case and matched better with observation.that of the in-situ data. The improvement in IOA was slightly less inas compared to that for temperature and winds.

566 5.2.3. High ozone episode after the passage of a front

In SETX, high ozone events during the fall season usually occurred after the passage of a cold
front (e.g., Rappenglück et al. 2008; Ngan and Byun, 2011; Ngan et al. 2012, Haman et al.
2014). Two factors may have contributed to the post-front ozone events: 1) followingFollowing a
cold spell-winds reverse direction and subsequent, light winds and sunny skies create an ideal
condition for ozone production and accumulation. 2) windWind reversal transportsmay transport
back the pollutants that were previously blown into the Gulf-previously, a phenomenon
commonly known as recirculation.

574 During the DISCOVER-AQ period, the two days with highest ozone <u>concentrations</u> were 575 09/25the 25th and the day 09/26 (26th of September as indicated in Figure 4), but the two days 576 exhibited different patterns. The 1-hour maximum ozone on 09/25the 25th was localized and 577 higher by about 40 ppb; than the 26th. In addition to heightened background ozone on the 25th, 578 the major contributor was the production resulting from favorable weather conditions: sunny, 579 overall light winds and shifting winds over the industrial area. The light morning land breeze
580	carried pollutants from ship channel area to the Galveston Bay. As the day warmed up, the bay
581	breeze started to develop and carry pollutants back to the land. This localized circulation was
582	described by Banta et al. (2005). Ngan et al. (2012) reported the same phenomenon in their
583	Texas Air Quality Study-II 2006 study. 09/26 is characterized by elevated background ozone
584	from early morning to late night.

Figure 8 shows the ozone time series of for the La Porte (C556), in Figure 1) site located in the
HSC area (Figure 1). In September, the . The highest hourly ozone of for September, 151 ppb
occurred at C556here at 13 CST of 09/25. From 9 CST to 12 CST, ozoneon the 25th. Ozone
rose from 10 ppb to 150 ppb. The large between 09-12 CST. Such a dramatic increase in ozone
was likely the result of ehemical productionincreased photochemical activity under favorable
meteorological conditionconditions in an area with accumulated precursors. Figures 9 and 10
depict the wind and ozone concentrations at 08 CST and 13 CST.

592 From the The wind plots of Figure 9, we can see indicate that the winds in the HGB region at 8 593 CST were light northerly for sites located on the north side while windsthey were mostly 594 westerly for the sites in the middle and south. The base case winds were all northerly while the 595 OA case had northwest winds for north side and west winds for the middle and south. TheHence, 596 the model winds in OA case were muchare more realistic than the winds in base case. The 909 597 CST winds were similar to those of 808 CST. As a result, the ozone statistics in Table 5 showed 598 that the OA case had much better correlation and IOA than the base case during 8 908-09 CST. 599 This example demonstrated OA'sthe ability of obs-nudging to correct erroneous winds. 600 However, later events showed OAit may not always be able to perform consistently.

601 The bay breeze started to develop at 10 CST near <u>the C556 site</u>. The early onset was likely to be

for a related to the warming up on the previous afternoon on 09/24 (as indicated in Figure 3). At 10

603 CST most other sites to the west of HSC experienced light northwest winds while windsthose at

604 HSC were<u>originated</u> from the northeast. Combined with the easterly bay breeze, a convergence

- zone was formed just below C556, where emissions from the HSC area stalled and accumulated.
- 606 At 13 CST, the whole region had light winds and the bay breeze was well developed. The
- 607 highest ozone indeed appeared in C556 and its vicinity. The rapid increase of ozone
- 608 concentration for C556 between $\frac{909}{2}$ -13 CST is shown in Figure 8.



610 Figure 8. Ozone time series of La Porte (C556) between 09/24_00 to 09/28_00 CST of 2013.





Figure 9. Zoom-in ozone concentrations (right) and wind plots (left) at 09/25_0813 CST of
201325 September for "No-OA" (top) and "1Hr-OA" (bottom). Ozone observation is in small
circle; wind observation is indicated by an orange arrow. La Porte site C556 is labeled. The
valuenumerical range of right-side colour scale is 0 to 200 ppb. Higher value than 200 ppb has
the same colour as 200 ppb.



Figure 10. Zoom-in ozone concentrations (right) and wind plots (left) at 09/25_13 CST of 201325 September for "No-OA" (top) and "1Hr-OA" (bottom). Ozone observation is in small circle; wind observation is indicated by an orange arrow. La Porte site C556 is labeled. Bay breeze is shown in the orange oval.

The ozone measurements from aircraft P3-B provided a more complete picture for 09/25'sthe
ozone evolution on 09/25. During the day, P-3Bthe aircraft flew around the industrial area,
Galveston Bay, and Galveston Island for about 9 hours. Figures 11 and 12 showedplot the ozone
concentrations along aircraft tracks at 08 and 13 CST, with surface. Surface layer ozone from the

643	"No-OA" case is provided as background. The background was only intended as a for reference.
644	At 08 CST, ozone level of 60-80 ppb aloft was already observed at three locations (three loops in
645	Fig.11): Galveston Island, Smith Point and inner city. Another high of ~90 ppb could be seen
646	above the HSC area. Ozonesonde observations over HGB showed the aloft ozone aloft
647	concentrations were normally typically ~40-50 ppb (e. g., Li and Rappengluck 2014)-) at the
648	height level. The higher-than normal ozone aloft suggested a post-front ozone recirculation
649	condition. Such high ozone aloft might raise surface ozone level-as a growing PBL downwardly
650	mixed the air aloft with near surface air. At 13 CST, high ozone over 100 ppb was observed at
651	multiple locations. The highest <u>aloft</u> ozone <u>aloft,of</u> ~ 160 ppb, occurred southwest of Smith Point
652	in the Galveston Bay. Such level of ozone-high increase in ozone concentrations was likely the
653	result of active productionphotochemistry in the industrial zone and around Galveston Bay:
654	indicating a high level of precursor accumulation in the area.

2013-09-25_08 CST P3B Ozone with No-OA Surface





657 OA surface ozone.

655





Figure 12. Ozone along aircraft tracks at 09/25_13 CST of September 25th, overlaid upon model
"No-OA" surface ozone. Plumes can be seen as dark purple circles in Galveston Bay.

Figure 13 shows hourly ozone vertical profiles from 08 CST to 16 CST of on September 25th, 661 with ozone being displayed on the x-axis and height on the y-axis. One observation dot was 662 663 averaged over all the grid cells in the same model layer. For example, during 08 09 CST, aircraft flew passing 30 cells at model's 5th layer. The 5th layer had a mid layer height of 287.5 m. The 664 averaged ozone of the 30 cells was 56 ppb. It should be noted that the The observed ozone was 665 666 averaged over multiple measurements in the same model cell, such so that they could be properly 667 compared to model output. Next, both model and observed ozone values- were averaged over all the grid cells in the same model layer, such that one dot represents the average ozone of all the 668 669 cells in the same layer. The 08 and 09 CST profiles showed there was a high ozone layer with 670 average ozone of ~65 ppb, stretching from 450 m to 1200 m height. In comparison, all model runs had lower ozone in this layer. The model biases, as shown in Figure 14, were about -10 ppb 671 672 at 08 CST and grew to -20 ppb at 09 CST. The discrepancies large discrepancy between low 673 surface ozone and ozone aloft was unusual and may be explained by the earlier reversal arrival of windshigh ozone air mass aloft: winds at surface layer still showed a light northwesterly in the 674

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675	early morning while winds aloft already changed to southerly. The observed ozone rose
676	continuously in following hours yet model simulated ozone stagnated around 60 ppb from
677	surface up to 2000 m until 15 CST. At 16 CST, the ozone of OA case in the lowermost (0-1 km)
678	layer rose 20 ppb over the previous hours yet the base case ozone increased only a few ppb.
679	Although different in magnitude, ozonethe aloft ozone had a few similar features to the surface
680	ozone. Firstly, the model missed the observed high ozone in the afternoon by a large margin. For
681	example, the base case underpredicted the 0-1 km level ozone by up to 50 ppb. The primary
682	cause for the lower ozone production was likely model's wind fields as both model and
683	observationobservations had a clear sky in industrial area and Galveston Bay. Secondly, nudging
684	clearly helped reducing the ozone biases aloft. In most plots of Figure 14, the OA case had lower
685	biases than the base case. The largest difference was at $16 \text{ CST}_{\overline{2}}$ when nudging reduced biases
686	from ~45 ppb to ~30 ppb in the $300 - 1000$ m layer.
687	While it is easy to understand the improvements in temperature and winds after obs-nudging was
007	while it is easy to understand the improvements in temperature and whiles after obs-hudging was
688	applied, it is more difficult to explain how other variables such as precipitation and clouds
689	reacted to obs-nudging. The indirect impact of these meteorological variables on ozone was
690	harder to assess. In our study, we did not evaluate clouds quantitatively as there were no
691	digitized cloud fraction data available for our modeling domains. A preliminary analysis on
692	convection showed that there were occasions in which model missed the convection or
693	precipitation and there were other occasions in which model created artificial convection. The

694 <u>convection cells were usually visible as "star-burst" from surface wind vector plots – arrows</u>

695 going out to different directions from a center. However, the mismatch in convection appeared to

696 <u>be not a serious issue since only a few occurrences were observed in the month of September.</u>



Figure 13. Vertical ozone profiles from 09/25_08 CST to 09/25_16 CST of 2013 for two cases
of No-OA and 1Hr-OA compared with corresponding observations.



Figure 14. Model vertical ozone biases from 09/25_08 CST to 09/25_16 CST of 2013 for two
cases of No-OA and 1Hr-OA.

704 6. Conclusions and Discussions

705 In this study, we performed two Weather Research and Forecasting (WRF) and Community 706 Multiscale Air Quality (CMAQ) model simulations to explore model sensitivity to observation 707 nudging. In evaluating meteorological and ozone conditions, we found that objective analysis 708 (OA)obs-nudging improved the meteorology and ozone performance as shown in the index of 709 agreement (IOA) of temperature, winds, and ozone. While the base case winds were overall well 710 simulated, observation obs-nudging significantly reduced the high wind biases (especially the 711 meridional wind) shown in the base case. For planetary boundary layer height, OAobs-nudging 712 reduced high biases in both daily maximum and daily minimum values. In the end, the combined 713 changes in meteorology lowered the ozone biases by about 3 ppb, a 35% reduction. There were 714 short time periods (such as between 07 and 09 CST on 09/25) when the simulated base case 715 model winds differ greatly-differed significantly from observation bservational data and OAobs-716 nudging significantly corrected the meteorological simulation problems, leading to much better ozone simulation. It should be noted that the However, model ozone biases are also impacted by 717 the emissions and model lateral boundary conditions. 718

719 While it is easy to understand the improvements in temperature and winds after OA was applied, more difficult to explain how other variables such as PBL and clouds reacted to OA. The 720 721 indirect impact of these meteorological variables on ozone was harder to assess. In our study, we 722 did not evaluate clouds quantitatively as there were no digitized cloud fraction data available for 723 our modeling domains. A preliminary analysis on convection showed that there were occasions in which model missed the convection or precipitation and there were other occasions in which 724 725 model created artificial convection. The convection cells were usually visible as "star burst" from surface wind vector plots - arrows going out to different directions from a center. However, 726 the mismatch in convection appeared to be not a serious issue since only a few occurrences were 727 728 observed in the month of September.

The only high ozone episode in the simulation period was related to the cold front passage. The
small-scale winds and high <u>aloft ozone aloft concentrations on 09/25</u>, likely contributed to the
ozone exceedance in the area. It is also possible that an unreported emission event played a role.
Since the maximum surface ozone at La Porte was much higher than the morning-time <u>aloft</u>
ozone <u>aloft</u>, the active local ozone production was likely the dominant factor. Analyses of <u>aloft</u>

734	ozone aloft on 09/25 showed while there was high aloft ozone aloft and large negative model
735	biases, the OA case tended to have smaller biases, especially in late hours.
736	Small-scale meteorological events are frequently cited for their contributions to high ozone
737	events. Model's capability in reproducing these events is critical in simulating such high ozone
738	episodes. The base case did not recreate the 09/25 September small-scale events likely due to
739	the complex winds and a lack of local information which can be used to steer model state closer
740	to reality. On the other hand, the inability of the OAsensitivity case to replicate the local winds is
741	likely a result of the imperfection of the nudging process which requirespending further
742	investigation. An ongoing study by the current authors suggests that errors in the metrological
743	fields from the default grid nudging files are important sources. Methods are being tested to
744	improve the quality of grid nudging files. Early results showed that the bay breeze which caused
745	the wind reversal around La Porte was well captured through improved grid nudging files. In
746	addition, more observational data (e.g., more sites and higher data frequency) and more testing
747	on the combination of OAnudging setting should help improve the OAobs-nudging performance.
748	Also, the impact of obs-nudging on precipitation and clouds should be further investigated to
749	understand their chain effect on chemistry.

750 Acknowledgement

751 **Acknowledgements**

- 752 The authors thank Texas Air Research Center (TARC) for its support through grant number
- 753 413UHH0144A and Air Quality Research Program (AQRP) through 14-014, the DISCOVER-
- 754 AQ team for the aircraft data, Vanessa Caicedo for LIDAR data, and the TCEQ CAMS site team
- 755 for the in-situ ozone and meteorological data.

756 References

- 757 Banta, R. M., Senff, C. J., White, A. B., Trainer, M., McNider, R. T., Valente, R. J., Mayor, S.
- D., Alvarez, R. J., Hardesty, R. M., and Parrish, D.: Daytime buildup and nighttime transport of
- virban ozone in the boundary layer during a stagnation episode, Journal of Geophysical Research:
 Atmospheres (1984–2012), J. Geophys. Res.-Atmos., 103, 22519-22544, 1998.
- 761 Banta, R. M., Senff, C. J., Nielsen-Gammon, J., Darby, L. S., Ryerson, T. B., Alvarez, R. J.,
- 762 Sandberg, S. R., Williams, E. J., and Trainer, M.: A bad air day in Houston, Bulletin of the
- 763 American Meteorological Society, B. Am. Meteorol. Soc., 86, 657-669, doi:10.1175/bams-86-5 764 657, 2005.
- 765 Banta, R. M., Senff, C. J., Alvarez, R. J., Langford, A. O., Parrish, D. D., Trainer, M. K., Darby,
- 766 L. S., Hardesty, R. M., Lambeth, B., Neuman, J. A., Angevine, W. M., Nielsen-Gammon, J.,
- 767 Sandberg, S. P., and White, A. B.: Dependence of daily peak O₃ concentrations near Houston,
- 768 Texas on environmental factors: Wind speed, temperature, and boundary-layer depth, Atmos.
- 769 Environ_{5.1} 45, 162-173, <u>doi:</u>10.1016/j.atmosenv.2010.09.030, 2011.
- 770 Byun, D., and Schere, K. L.: Review of the governing equations, computational algorithms, and

other components of the models-3 Community Multiscale Air Quality (CMAQ) modeling

system, Applied Mechanics Reviews, Appl. Mech. Rev., 59, 51-77, doi:10.1115/1.2128636,
2006.

- 774 Byun, D., Ngan, F., Li, X., Lee, D., Kim, S..: "Analysis of Air Pollution Events in Summer 2006
- 775 and Preparation of Model Input Data for the Assessment Study", Grant No. 582-5-64594- FY07-
- 776 02, Final Report: Evaluation of Retrospective MM5 and CMAQ Simulations of TexAQS-II
- Period with CAMS Measurements, Texas Commission on Environmental Quality, February,
- 778 2008, 25 pp
- 779 Cheng, F.Y., and Byun, D. : Application of high resolution land use and land cover data for
- 780 atmospheric modeling in the Houston-Galveston metropolitan area, Part I: Meteorological
- 781 simulation results, Atmos. Env., 42, 7795-7811, <u>doi:</u>10.1016/j.atmosenv.2008.04.055, 2008.
- 782 Cheung, V. T., and Wang, T.: Observational study of ozone pollution at a rural site in the
- 783 Yangtze Delta of China, Atmos. Env., 35, 4947-4958, 2001.

784	Choi, Y.: The impact of satellite-adjusted NO_x emissions on simulated NO_x and O_3 discrepancies
785	in the urban and outflow areas of the Pacific and Lower Middle US, Atmos. Chem. Phys., 14,
786	<u>675-690, doi:10.5194/acp-14-675-2014, 2014.</u>
787	Choi, Y. and Souri, A.: Chemical condition and surface ozone in large cities of Texas during the
788	last decade: observational evidence from OMI, CAMS, and Model Analysis, Remote Sensing of
789	Environ., 168:90-101, doi:10.1016/j.rse.2015.06.026, 2015
790	Choi, Y., Kim, H., Tong, D., and Lee, P.: Summertime weekly cycles of observed and modeled
791	NO _x and O ₃ concentrations as a function of satellite-derived ozone production sensitivity and
792	land use types over the Continental United States, Atmos. Chem. Phys., 12, 6291-6307,
793	<u>doi:10.5194/acp-12-6291-2012</u> , 2012
794	Choi, Y.: The impact of satellite-adjusted NO _x -emissions on simulated NO _x -and O ₃ -discrepancies
795	in the urban and outflow areas of the Pacific and Lower Middle US, Atmos Chem Phys, 14, 675-
796	690, DOI: 10.5194/acp 14-675-2014, 2014.
797	Choi, Y. and Souri, A.: Chemical condition and surface ozone in large cities of Texas during the
797 798	Choi, Y. and Souri, A.: Chemical condition and surface ozone in large cities of Texas during the last decade: observational evidence from OMI, CAMS, and Model Analysis, Remote Sensing of
798	last decade: observational evidence from OMI, CAMS, and Model Analysis, Remote Sensing of
798 799	last decade: observational evidence from OMI, CAMS, and Model Analysis, Remote Sensing of Environment, 168:90-101, DOI:10.1016/j.rse.2015.06.026, 2015
798 799 800	last decade: observational evidence from OMI, CAMS, and Model Analysis, Remote Sensing ofEnvironment, 168:90-101, DOI:10.1016/j.rse.2015.06.026, 2015Cuchiara, G. C., Li, X., Carvalho, J., and Rappenglück, B.: Intercomparison of planetary
798 799 800 801	 last decade: observational evidence from OMI, CAMS, and Model Analysis, Remote Sensing of Environment, 168:90-101, DOI:10.1016/j.rse.2015.06.026, 2015 Cuchiara, G. C., Li, X., Carvalho, J., and Rappenglück, B.: Intercomparison of planetary boundary layer parameterization and its impacts on surface ozone concentration in the
 798 799 800 801 802 	 last decade: observational evidence from OMI, CAMS, and Model Analysis, Remote Sensing of Environment, 168:90–101, DOI:10.1016/j.rse.2015.06.026, 2015 Cuchiara, G. C., Li, X., Carvalho, J., and Rappenglück, B.: Intercomparison of planetary boundary layer parameterization and its impacts on surface ozone concentration in the WRF/chem model for a case study in Houston, Texas, Atmos. Environ, 96, 175-185, doi:
 798 799 800 801 802 803 	 last decade: observational evidence from OMI, CAMS, and Model Analysis, Remote Sensing of Environment, 168:90-101, DOI:10.1016/j.rse.2015.06.026, 2015 Cuchiara, G. C., Li, X., Carvalho, J., and Rappenglück, B.: Intercomparison of planetary boundary layer parameterization and its impacts on surface ozone concentration in the WRF/chem model for a case study in Houston, Texas, Atmos. Environ, 96, 175-185, doi: 10.1016/j.atmosenv.2014.07.013, 2014.
 798 799 800 801 802 803 804 	 last decade: observational evidence from OMI, CAMS, and Model Analysis, Remote Sensing of Environment, 168:90–101, DOI:10.1016/j.rse.2015.06.026, 2015 Cuchiara, G. C., Li, X., Carvalho, J., and Rappenglück, B.: Intercomparison of planetary boundary layer parameterization and its impacts on surface ozone concentration in the WRF/chem model for a case study in Houston, Texas, Atmos. Environ, 96, 175-185, doi: 10.1016/j.atmosenv.2014.07.013, 2014. Czader, B. H., Li, X. S., and Rappenglueck, B.: CMAQ modeling and analysis of radicals,
 798 799 800 801 802 803 804 805 	 last decade: observational evidence from OMI, CAMS, and Model Analysis, Remote Sensing of Environment, 168:90-101, DOI:10.1016/j.rse.2015.06.026, 2015 Cuchiara, G. C., Li, X., Carvalho, J., and Rappenglück, B.: Intercomparison of planetary boundary layer parameterization and its impacts on surface ozone concentration in the WRF/chem model for a case study in Houston, Texas, Atmos₂ Environ, 96, 175-185, doi: 10.1016/j.atmosenv.2014.07.013, 2014. Czader, B. H., Li, X. S., and Rappenglueck, B.: CMAQ modeling and analysis of radicals, radical precursors, and chemical transformations, J₂ Geophys₂ Res-<u>-</u>Atmos₅. 118, 11376-11387,
 798 799 800 801 802 803 804 805 806 	 last decade: observational evidence from OMI, CAMS, and Model Analysis, Remote Sensing of Environment, 168:90-101, DOI:10.1016/j.rse.2015.06.026, 2015 Cuchiara, G. C., Li, X., Carvalho, J., and Rappenglück, B.: Intercomparison of planetary boundary layer parameterization and its impacts on surface ozone concentration in the WRF/chem model for a case study in Houston, Texas, Atmos, Environ, 96, 175-185, doi: 10.1016/j.atmosenv.2014.07.013, 2014. Czader, B. H., Li, X. S., and Rappenglueck, B.: CMAQ modeling and analysis of radicals, radical precursors, and chemical transformations, J, Geophys, Res-, Atmos, 118, 11376-11387, doi:-10.1002/Jgrd.50807, 2013.

- 810 Darby, L. S.: Cluster analysis of surface winds in Houston, Texas, and the impact of wind
- 811 patterns on ozone, Journal of Applied Meteorology, J. Appl. Meteorol., 44, 1788-1806, doi:
- 812 10.1175/jam2320.1, 2005.
- 813 Daum, P. H., Kleinman, L. I., Springston, S. R., Nunnermacker, L. J., Lee, Y. N., Weinstein-
- 814 Lloyd, J., Zheng, J., and Berkowitz, C. M.: Origin and properties of plumes of high ozone
- 815 observed during the Texas 2000 Air Quality Study (TexAQS 2000), J. Geophys. Res-.-Atmos.
 816 109, Artn-D17306, doi:-10.1029/2003jd004311, 2004.
- 817 Deng, A., Stauffer, D., Gaudet, B., Dudhia, J., Hacker, J., Bruyere, C., Wu, W., Vandenberghe,
- 818 F., Liu, Y., Bourgeois, A.: Update on WRF-ARW End-to-end Multi-scale FDDA System. 10th
- 819 WRF Users' Workshop, Boulder, CO, June 23, 2009. [Available online at,
- 820 http://www2.mmm.ucar.edu/wrf/users/workshops/WS2009/abstracts/1-09.pdf]NCAR, 2009
- 821 Foley, K. M., Roselle, S. J., Appel, K. W., Bhave, P. V., Pleim, J. E., Otte, T. L., Mathur, R.,
- 822 Sarwar, G., Young, J. O., Gilliam, R. C., Nolte, C. G., Kelly, J. T., Gilliland, A. B., and Bash, J.
- 823 O.: Incremental testing of the Community Multiscale Air Quality (CMAQ) modeling system
- 824 version 4.7, GeoscientificGeosc. Model Development, 3, 205-226, doi:10.5194/gmd-3-205-2010,
 825 2010.
- 826 Gilliam, R. C., and Pleim, J. E.: Performance Assessment of New Land Surface and Planetary
- Boundary Layer Physics in the WRF-ARW, Journal of Applied Meteorology and Climatology, J.
 Appl. Meteorol. Climatol., 49, 760-774, doi:10.1175/2009jamc2126.1, 2010.
- 829 Haman, C.L., Lefer, B., Morris, G.A.: Seasonal Variability in the Diurnal Evolution of the
- 830 <u>Boundary Layer in a Near-Coastal Urban Environment. J. Atmos. Ocean Tech., 29, 697-710,</u>
 831 2012.
- 832 Haman, C.L., Couzo, E., Flynn, J.H., Vizuete, W., Heffron, B., Lefer, B.L.: Relationship
- 833 between boundary layer heights and growth rates with ground-level ozone in Houston, Texas. J
- 834 Geophys Res-Atmos., 119, 6230-6245, 2014.
- 835 Houyoux, M., Vukovich, J., Brandmeyer, J., 2000. Sparse Matrix Kernel Emissions Modeling
- 836 System: SMOKE User Manual. MCNC-North Carolina Supercomputing Center. Available at:
- 837 <u>https://cmascenter.org/smoke/.</u>

838	Haman, C.L., Lefer, B., Morris, G.A.: Seasonal Variability in the Diurnal Evolution of the
839	Boundary Layer in a Near-Coastal Urban EnvironmentJ Atmos Ocean Tech 29, 697-710, 2012.
840	Huang, XY., Xiao, Q., Barker, D. M., Zhang, X., Michalakes, J., Huang, W., Henderson, T.,
841	Bray, J., Chen, Y., and Ma, Z.: Four-dimensional variational data assimilation for WRF:
842	Formulation and preliminary results, MonthlyMon. Weather Review, Rev., 137, 299-314, 2009.
843	Kleinman, L. I., Daum, P. H., Lee, Y. N., Nunnermacker, L. J., Springston, S. R., Weinstein-
844	Lloyd, J., and Rudolph, J.: Ozone production efficiency in an urban area, J Geophys Res
845	Atmos, 107, Artn-4733, doi:-10.1029/2002jd002529, 2002.
846	Le Dimet, F.X. and Talagrand, O.: Variational algorithms for analysis and assimilation of
847	meteorological observations: theoretical aspects, Tellus, 38A, 97-110, 1986.
848	Lefer, B., Rappengluck, B., Flynn, J., and Haman, C.: Photochemical and meteorological
849	relationships during the Texas II Radical and Aerosol Measurement Project (TRAMP),
850	Atmospheric Environment, 44, 4005-4013, DOI 10.1016/j.atmosenv.2010.03.011, 2010.
851	Li, X., and Rappenglück, B.: A WRF-CMAQ study on spring time vertical ozone structure in
852	Southeast Texas, Atmos. Environ., 97, 363-385, doi:10.1016/j.atmosenv.2014.08.036, 2014.
853	Li, X., Lee, D., Kim, ST., Kim, H., Ngan, F., Cheng, F., and Byun, D.: Performance
854	Evaluation of a Year-long Run of an Air Quality Forecasting System for Southeast Texas,
855	10th Conference on Atmospheric Chemistry, New Orleans, January 2008, 2008. [Available
856	online at, http://ams.confex.com/ams/pdfpapers/134453.pdf]
857	Li, X., and Rappenglück, B.: A WRF-CMAQ study on spring time vertical ozone structure in
858	Southeast Texas, Atmospheric Environment, 97, 363–385, doi: 10.1016/j.atmosenv.2014.08.036,
859	2014.
860	Liu, Y., Bourgeois, A., Warner, T., Swerdlin, S., and Hacker, J.: An implementation of
861	observation nudging-based FDDA into WRF for supporting ATEC test operations, 2005 WRF
862	user workshop, Boulder, CO, 2005. [Available online at,
863	http://www2.mmm.ucar.edu/wrf/users/workshops/WS2005/abstracts/Session10/7-Liu.pdf]

864	Liu, Y., Bourgeois, A., Warner, T., Swerdlin, S., and Yu, W.: An update on "observation
865	nudging"-based FDDA for WRF-ARW: Verification using OSSE and performance of real-time
866	forecasts, 2006 WRF user workshop, Boulder, CO, 2006. [Available online at,
867	http://www2.mmm.ucar.edu/wrf/users/workshops/WS2006/abstracts/Session04/4_7_Liu.pdf]
868	Ngan, F., and Byun, D.: Classification of Weather Patterns and Associated Trajectories of High-
869	Ozone Episodes in the Houston Galveston Brazoria Area during the 2005/06 TexAQS II,
870	Journal of Applied Meteorology and Climatology, 50, 485–499, DOI: 10.1175/2010jamc2483.1,
871	2011.
872	Ngan, F., Byun, D., Kim, H., Lee, D., Rappengluck, B., and Pour-Biazar, A.: Performance
873	assessment of retrospective meteorological inputs for use in air quality modeling during TexAQS
874	2006, Atmos. Environ, 54, 86-96, DOI: doi: 10.1016/j.atmosenv.2012.01.035, 2012.
875	Mason, R.; Strum, M.; Houyoux, M. Technical Support Document (TSD) Preparation of
876	Emissions Inventories for the Version 4, 2005-based Platform; U.S. Environmental Protection
877	Agency, Office of Air and Radiation, Office of Air Quality Planning and Standards, Air Quality
878	Assessment Division, 2010
879	Olaguer, E. P., Rappengluck, B., Lefer, B., Stutz, J., Dibb, J., Griffin, R., Brune, W. H., Shauck,
880	M., Buhr, M., Jeffries, H., Vizuete, W., and Pinto, J. P.: Deciphering the Role of Radical
881	Precursors during the Second Texas Air Quality Study, Journal of the J. Air & Waste
882	Management Association, Manag. Assoc., 59, 1258-1277, doi:-10.3155/1047-3289.59.11.1258,
883	2009.
884	Otte, T. L.: The impact of nudging in the meteorological model for retrospective air quality
885	simulations. Part I: Evaluation against national observation networks, Journal of J. Applied
886	Meteorology and Climatology, Meteorol. Climatol., 47, 1853-1867,
887	doi:10.1175/2007jamc1790.1, 2008.
888	Pan, S., Choi, Y., Roy, A., Li, X., Jeon, W., and Souri, A.: Modeling the uncertainty of several
889	VOC and ints impact on simulated VOC and ozone in Houston, Texas, Atmos. Environ, 120,

 889
 404-416, 2015

- 891 Parrish, D. D., Allen, D. T., Bates, T. S., Estes, M., Fehsenfeld, F. C., Feingold, G., Ferrare, R.,
- 892 Hardesty, R. M., Meagher, J. F., Nielsen-Gammon, J. W., Pierce, R. B., Ryerson, T. B., Seinfeld,
- 893 J. H., and Williams, E. J.: Overview of the Second Texas Air Quality Study (TexAQS II) and the
- 894 Gulf of Mexico Atmospheric Composition and Climate Study (GoMACCS), J. Geophys. Res-.-
- 895 Atmos,... 114, D00f13, doi:-10.1029/2009jd011842, 2009.
- 896 Pour-Biazar, A., McNider, R. T., Roselle, S. J., Suggs, R., Jedlovec, G., Byun, D. W., Kim, S.,
- 897 Lin, C. J., Ho, T. C., Haines, S., Dornblaser, B., and Cameron, R.: Correcting photolysis rates on

898 the basis of satellite observed clouds, J. Geophys. Res-.-Atmos, 112, D10302, doi:

- 899 10.1029/2006jd007422, 2007.
- 900 Rappengluck, B., Perna, R., Zhong, S. Y., and Morris, G. A.: An analysis of the vertical structure
- 901 of the atmosphere and the upper-level meteorology and their impact on surface ozone levels in
- 902 Houston, Texas, J. Geophys. Res... 113, D17315, doi:-10.1029/2007jd009745, 2008.
- 903 Rappenglück, B., Lefer, B., Mellqvist, J., Czader, B., Golovko, J., Li, X., Alvarez, S., Haman,
- 904 C., and Johansson, J., 2011: University of Houston Study of Houston Atmospheric Radical
- Precursors (SHARP), Report to the Texas Commission on Environmental Quality, August 2011,
 145 pp
- 907 Skamarock, W. C., Klemp, J. B., Dudhia, J., Gill, D. O., Barker, M., Duda, K. G., Huang, Y.,
- Wang, W., and Powers, J. G.: A description of the Advanced Research WRF Version 3, 1-113,2008.
- 910 Stauffer, D. R., and Seaman, N. L.: Use of 4-dimensional data assimilation in a limited-area
- 911 mesoscale model .1. Experiments with synoptic-scale data, <u>MonthlyMon.</u> Weather <u>Review,Rev.</u>
- 912 118, 1250-1277, 1990.
- Stauffer, D. R., and Seaman, N. L.: Multiscale 4-dimensional data assimilation, Journal of
 Applied Meteorology, J. Appl. Meteorol., 33, 416-434, 1994.
- 915 Tucker, S. C., R. M. Banta, A. O. Langford, C. J. Senff, W. A. Brewer, EJ. Williams, B. M.
- 916 Lerner, H. D. Osthoff, and R. M. Hardesty. : Relationships of coastal nocturnal boundary layer
- 917 winds and turbulence to Houston ozone concentrations during TexAQS 2006. Journal of
- 918 Geophysical Research: Atmospheres, 115, <u>D10304, doi: 10.1029/2009JD013169no. D10</u> (2010).

- 919 Willmott, C. J.: On the Validation of Models, Physical Geography, 2, 184-194,
- 920 <u>doi:10.1080/02723646.1981.10642213, 1981.</u>
- 921 Zhong, S. Y., In, H. J., and Clements, C.: Impact of turbulence, land surface, and radiation
- 922 parameterizations on simulated boundary layer properties in a coastal environment, J. Geophys.
- 923 Res-<u>.-</u>Atmos,.. 112, D13110, doi:-10.1029/2006jd008274, 2007.
- 924

- 926 Table 1. Major WRF physics and FDDA Options, the numbers in the parentheses are the related
- 927 settings in WRF namelist file.

V3.5.1
Lin et al Scheme (2)
RRTMG (4)
New Goddard scheme (5)
Monin-Obukhov with CB viscous sublayer scheme (1)
Unified Noah LSM (2)
None
YSU (1)
Kain-Fritsch (1)
Grid nudging on for all; Observation-nudging on for the OA case

- 931 Table 2. Major CMAQ Options, the text in the parentheses are the related settings in CMAQ
- 932 build script.

CMAQ version	V5.0.1
Chemical Mechanism	cb05tucl_ae5_aq: CB05 gas-phase mechanism with active
	chlorine chemistry, updated toluene mechanism, fifth-generation
	CMAQ aerosol mechanism with sea salt, aqueous/cloud
	chemistry
Lightning NOx emission	Included by using inline code
Horizontal advection	YAMO (Yamartino) (hyamo)
Vertical advection	WRF omega formula (vwrf)
Horizontal	Multiscale (multiscale)
mixing/diffusion	
Vertical mixing/diffusion	Asymmetric Convective Model (ACM) version 2 (acm2)
Chemistry solver	EBI (Euler Backward Iterative) (ebi_cb05tucl)
Aerosol	AERO5 for sea salt and thermodynamics (aero5)
Cloud Option	ACM cloud processor for AERO5 (cloud_acm_ac5)
Boundary conditions	Default static profiles

Table 3 Statistics of surface T, U-wind and V-wind for three WRF simulations: N - data points;

- Corr - Correlation; IOA - Index of Agreement; RMSE - Root Mean Square Error; MAE - Mean
- Absolute Error; MB Mean Bias; O Observation; M Model; O_M Observed Mean; M_M -
- Model Mean; SD - Standard Deviation; Units for RMSE/MAE/MB/O_M/M_M/O_SD/M_SD: degree C

945

		_								
Surface temperature T										
Case	Ν	Corr	IOA	RMSE	MAE	MB	O_M	M_M	O_SD	M_SD
No-OA	41058	0.83	0.89	2.0	1.5	0.9	27.4	28.3	3.1	2.8
1Hr-OA	41058	0.94	0.97	1.0	0.8	0.0	27.4	27.4	3.1	3.1
Surface U wind										
Case	N	Corr	IOA	RMSE	MAE	MB	O_M	M_M	O_SD	M_SD
No-OA	43246	0.76	0.84	1.4	1.1	-0.6	-1.3	-1.9	1.6	1.9
1Hr-OA	43246	0.81	0.89	1.0	0.8	-0.3	-1.3	-1.6	1.6	1.6
Surface V wind										
Case	N	Corr	IOA	RMSE	MAE	MB	O_M	M_M	O_SD	M_SD
No-OA	43246	0.76	0.8	2.1	1.7	1.2	0.4	1.7	2.0	2.6
1Hr-OA	43246	0.80	0.89	1.2	0.9	-0.1	0.4	0.4	2.0	2.0

Table 4 Statistics of ozone for CMAQ simulations, see table 3 for column header information

Case	N	Corr	IOA	RMSE	MAE	MB	O_M	M_M	O_SD	M_SD
No-OA	33308	0.72	0.78	14.9	12.3	9.3	24.4	33.7	16.5	14.1
1Hr-OA	33308	0.73	0.83	13.8	11.0	6.1	24.4	30.6	16.5	17.4

951 agreement are unitless. The red numbers indicate the three hours (07 CST to 09 CST) when the

952 ozone in 1Hr-OA case is significantly better than the No-OA case due to much improved winds.

		1			
		No-OA		1Hr-OA	
	Ν	Corr	IOA	Corr	IOA
Hr All	1150	0.79	0.86	0.81	0.88
0	48	0.04	0.30	0.40	0.46
1	43	0.20	0.24	0.36	0.30
2	48	0.14	0.25	0.35	0.35
3	48	0.19	0.30	0.32	0.35
4	48	0.27	0.36	0.31	0.35
5	47	0.24	0.36	0.28	0.37
6	47	0.33	0.38	0.35	0.37
7	48	0.06	0.39	0.29	0.47
8	48	0.09	0.43	0.53	0.63
9	47	0.05	0.41	0.55	0.74
10	47	-0.10	0.29	0.30	0.51
11	47	0.13	0.39	-0.07	0.36
12	49	0.09	0.38	0.25	0.40
13	49	-0.09	0.37	0.36	0.46

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