

## P1.8 SENSITIVITY OF WRF/CHEM PREDICTIONS TO METEOROLOGICAL SCHEMES

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### 1. INTRODUCTION

Three-dimensional (3-D) atmospheric models are an important tool in studying meteorological variables and their impacts on chemical species. Understanding the effects of meteorological parameterizations on meteorological and chemical predictions of 3-D models is necessary to the use of appropriate parameterizations for simulating different episodes. Performance of these schemes may vary from episode to episode, and may be highly dependent upon the horizontal grid spacing and time resolution used. The accuracy of predicted meteorological variables (e.g., temperature (T), relative humidity (RH), wind speed and direction, and planetary boundary layer (PBL) height) is a prerequisite for accurate predictions of chemical species and depends on the performance of meteorological parameterizations.

Operating online, the Weather Research and Forecasting/Chemistry (WRF/Chem) model uses the state of the science WRF meteorological model and integrates the simulation of chemical species formation and transport simultaneously with meteorological predictions. The objectives of this study are to examine the simulated meteorological variables using different combinations of land-surface models (LSMs) and planetary boundary layer (PBL) parameterizations, and the effects of the simulated meteorology on chemical predictions.

WRF/Chem offers several options for meteorological physics, as well as several gas-phase mechanisms, and aerosol modules. Two options critical to WRF/Chem simulations are the LSM and PBL schemes. This study evaluates several well known meteorological schemes including two LSMs (the Noah (Ek et al., 2003) and slab soil (Dudhia, 1996) and two PBL schemes (the Yonsei University (YSU) (Hong and Dudhia, 2003) and Mellor-Yamada-Janjic (Mellor and Yamada, 1982; Janjic, 1994) (MYJ) PBL

schemes). The Noah LSM is a popular scheme developed jointly by the National Center for Environmental Prediction, Oregon State University, the Air Force, and the Hydrologic Research Lab. The five-layer slab soil model provides an improvement to the two-layer scheme used in MM5. Both the YSU and MYJ PBL schemes are turbulence parameterizations commonly employed in several operational meteorological models. The MYJ is a PBL scheme that explicitly defines turbulent quantities, using a conservation of turbulent kinetic energy (TKE). The YSU scheme, however, is a modification of the Medium Range Forecast (MRF) PBL scheme, using an explicit treatment of entrainment rather than implicit. The MYJ scheme defines the PBL height using TKE, while the YSU scheme is dependent upon the bulk Richardson number.

### 2. DOMAIN AND MODEL CONFIGURATION

WRF/Chem is applied with a 12-km horizontal grid spacing for a 5-day (28 August through 2 September 2000) episode from the Texas Air

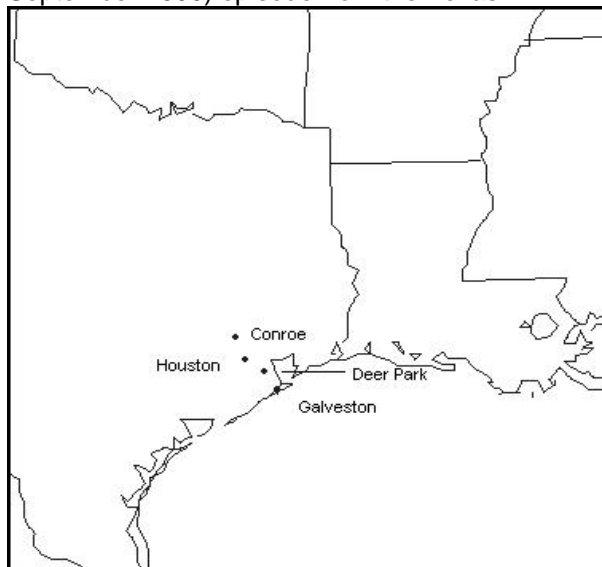


Figure 1. Map of WRF/Chem simulation domain for 28 August - 2 September 2000 TexAQS-2000 episode.

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Quality Study 2000 (TexAQS-2000) in the southern U.S. The domain for this study centers on the Houston-Galveston area in eastern Texas, as shown in Figure 1.

The chemistry used in this study consists of the Second Generation Regional Acid Deposition Model (RADM2) gas-phase mechanism (Stockwell et al., 1990), the Modal Aerosol Dynamics Model (MADE) with the Secondary Organic Aerosol Model (SORGAM)(Ackermann et al., 1998; Schell et al., 2001).

To examine the effects of different meteorological schemes on chemical simulations, a base simulation with the Noah LSM coupled with the YSU PBL scheme (hereafter referred to as SOR\_NOAH) is conducted to provide a benchmark to two sensitivity simulations in which

different combinations of meteorological schemes are used. The slab LSM and the YSU PBL scheme are used in the first sensitivity simulation and the Noah LSM and MYJ PBL scheme are used in the second sensitivity simulation (hereafter referred to as SOR\_SLAB and SOR\_TKE, respectively).

### 3. MODEL EVALUATION

#### 3.1 Meteorological Variables

The meteorological predictions of WRF/Chem are compared against observations from TexAQS-2000. The analysis consists mainly of temporal and statistical analysis. Figure 2 shows the temporal distributions of temperature, wind speed and direction, and RH at Deer Park, Texas and PBL height at LaMarque, Texas, a location with observations near Deer Park. The predicted meteorological variables correspond well with observed values for the first two days at most locations, especially in the case of temperature and RH. During days three through five, model deviations grow significantly away from observed values for all simulations. This can be attributed to the lack of a data assimilation system in the version of WRF used here, which is currently under development. In general, each model simulation underpredicts the daily maximum temperature. The SOR\_NOAH simulation overpredicts nighttime temperatures at most locations, while SOR\_SLAB shows a general underprediction at nearly all times during days three through five. The SOR\_SLAB simulation also gives RH more accurately than SOR\_TKE and SOR\_NOAH, as the latter two grossly underpredict nighttime values by up to 58% of the observed value during days 3-5. All three simulations reproduce the observed wind speed relatively well at all locations and during most hours. Overall, wind direction simulations showed significant deviations from observed values during nighttime hours. However, one limitation of evaluating wind directions is that because wind is a vector, the numeric values of differences between the observations and predictions may not reflect well the actual differences in wind rose plots when the numeric differences are greater than 180°. For example, at 6:00AM LST on August 29 at Deer Park, the observed wind direction is 11°, and the predicted wind directions from the three model runs range from 228.3° to 254.1°. The numerical differences range from 217.3° to 243.1°, whereas the actual differences on wind rose plots range from 142.7°

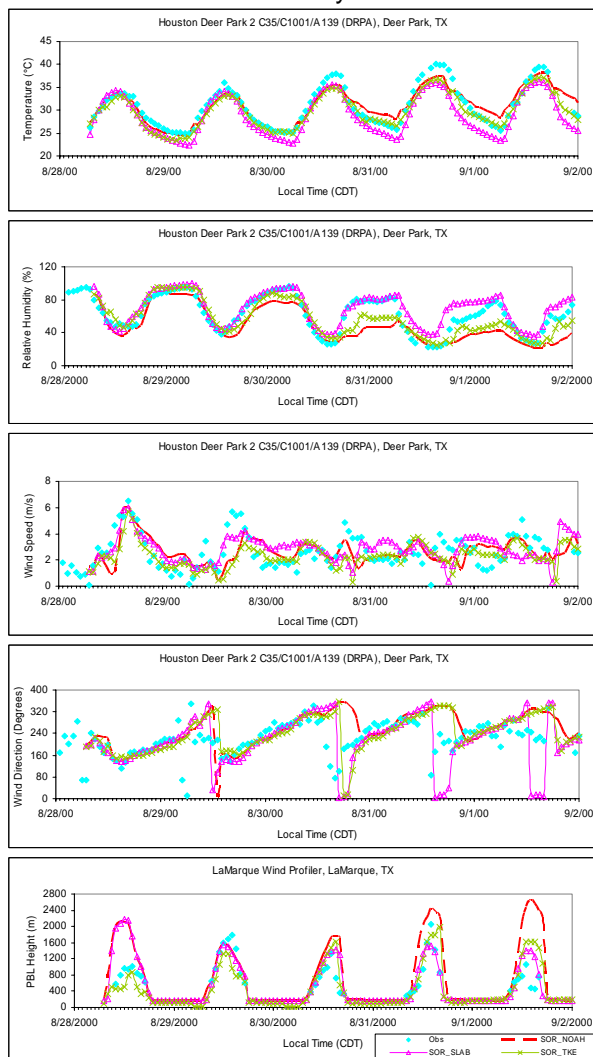


Figure 2. Temporal distributions of (a) T, (b) RH, (c) wind speed, and (d) wind direction at Deer Park, Texas; and (e) PBL height at LaMarque, Texas during 28 August - 2 September, 2000.

Table 1. Performance statistics for T at 2-meter height (T2), RH, PBL height (1A) and wind speed and direction (1B).

1A	T2			RH			PBL Height		
	NOAH	SLAB	TKE	NOAH	SLAB	TKE	NOAH	SLAB	TKE
MeanObs	31.2	31.2	31.2	62.5	62.5	62.5	1089.3	1089.3	1089.3
MeanMod	30.9	28.9	29.9	45.2	66.2	52.6	1519.4	1288.6	913.4
Number	3502	3502	3502	795	795	795	203	203	203
Corr	.88	.91	.91	.70	.75	.74	.66	.77	.64
NMB (%)	-1	-7	-4	-28	6	-16	39	18	-16
NME (%)	5	8	6	31	20	23	45	29	30
1B	WSP			WDR					
	NOAH	SLAB	TKE	NOAH	SLAB	TKE			
MeanObs	2.9	2.9	2.9	210	210	210			
MeanMod	3	2.9	2.6	228.5	223.7	223.7			
Number	3502	3502	3502	3391	3391	3391			
Corr	0.35	0.30	0.47	0.46	0.40	0.46			
NMB (%)	3	-1	-12	9	7	7			
NME (%)	0.42	0.42	0.37	0.21	0.23	0.21			

to 116.9°, respectively. For PBL height predictions at most locations, both SOR\_NOAH and SOR\_SLAB overpredict values for day one at all locations. During days two through five, SOR\_TKE and SOR\_SLAB give a closer agreement to the observed PBL heights, with SOR\_TKE showing a slight overprediction and SOR\_NOAH showing consistent overprediction during most time periods. All simulations, however, represent the early development of the PBL height relatively well at all locations. Temporal analyses also show that model discrepancies grow significantly larger during days three through five, mainly because of the lack of a data assimilation system in WRF, as mentioned previously.

Several statistical metrics suggested by Yu et al. (2003) are calculated to evaluate overall model performance. Tables 1A and 1B show several statistical parameters for meteorological predictions. In terms of normalized mean bias (NMB), the SOR\_SLAB simulation gives better performance than SOR\_NOAH and SOR\_TKE for RH (5.8% vs. -27.8% and -15.9%) and wind speed (-0.5% vs. 2.6% and -12.2%). Both SOR\_TKE and SOR\_SLAB slightly outperform SOR\_NOAH for wind direction (both 6.5% vs. 8.8%). Statistically, SOR\_TKE predicts PBL height more accurately than SOR\_NOAH and SOR\_SLAB (-16.1% vs. 39.5% and 18.3%). The general underprediction of the MYJ scheme is generally consistent with results from past studies (e.g., Fast, 2005). Due to the implicit nature of the boundary layer depth in the YSU scheme, errors within variables used to calculate vertical motions can cause errors in simulated depths, whereas the MYJ scheme (which uses more complete physics)

seems to have less bias (Alapaty et al., 1996). However, as shown in Figure 2, SOR\_SLAB may seem to outperform SOR\_NOAH and SOR\_TKE during days two through five, but the magnitude of the overprediction on day one dominates the overall statistical performance of the simulation. Because SOR\_TKE is relatively closer to observations at all days, it shows less bias than SOR\_SLAB, even though SOR\_SLAB performs better on Days 4-5. SOR\_NOAH, though, outperforms SOR\_SLAB and SOR\_TKE at forecasting temperature (-1.1% vs. -7.4% and -4.1%).

### 3.2 Chemical Species

The model chemical predictions are also evaluated using observed chemical data taken from the TexAQS-2000. The mixing ratios of O<sub>3</sub> and CO, as well as the concentrations of fine particulate matter (PM<sub>2.5</sub>) and sulfate (SO<sub>4</sub><sup>2-</sup>) are the initial foci of our analyses. Temporal and statistical analyses are conducted (where data were available), along with an intercomparison among three sets of model results.

Figure 3 shows the temporal distribution of chemical species at Deer Park, Texas. For O<sub>3</sub>, all three model simulations predict daytime concentrations relatively well, with the exception of some sites (as shown in Figure 3 at Deer Park) where deviations are 41-52% less than the maximum observed values on days 3-4. Each simulation significantly overpredicts nighttime concentrations, by as much as 200% of the observed value in some cases. Of the three simulations, however, SOR\_TKE more adequately replicates nighttime O<sub>3</sub>, with the lowest NMB.

Analyses of the CO plot shows that SOR\_TKE gives mixing ratios that are closer to the observations than the other two simulations. One possible explanation for this is that SOR\_TKE gives lower PBL heights on days 1-3 than SOR\_NOAH and SOR\_SLAB, which allows less volume for CO to mix out. Large model deviations exist for CO predictions during nighttime. At night, the MYJ scheme simulates a shallower boundary layer, which would explain the higher predictions of CO during nighttime by SOR\_TKE. For PM<sub>2.5</sub> and SO<sub>4</sub><sup>2-</sup>, the results from the three

model simulations are quite similar during most time periods. At most locations, SO<sub>2</sub> concentrations predicted by SOR\_SLAB are greater than those predicted by either SOR\_NOAH or SOR\_TKE, which leads to higher predicted SO<sub>4</sub><sup>2-</sup> concentrations. Relatively minor differences among model simulations exist for PM<sub>2.5</sub> with the exception of higher simulated daily maximum values by the SOR\_SLAB simulation.

Table 2. Performance statistics for predictions of O<sub>3</sub>, CO, NO, NO<sub>2</sub>, and PM<sub>2.5</sub>.

O <sub>3</sub>	Daytime			Nighttime			Daily		
	NOAH	SLAB	TKE	NOAH	SLAB	TKE	NOAH	SLAB	TKE
MeanObs	54.5	54.5	54.4	24.8	24.8	24.8	39.4	39.4	39.4
MeanMod	58.7	59.7	57.1	46.2	48.1	33.9	52.1	53.6	45
Number	3270	3270	3270	3139	3139	3139	6123	6123	6123
Corr	.62	.57	.63	.61	.54	.60	.70	.60	.70
NMB (%)	8	9	5	87	94	37	30	40	10
NME (%)	28	30	29	93	101	59	50	40	40
CO	Daytime			Nighttime			Daily		
	NOAH	SLAB	TKE	NOAH	SLAB	TKE	NOAH	SLAB	TKE
MeanObs	390.4	390.4	390.4	433.6	433.6	433.6	411.8	411.8	411.8
MeanMod	257.8	309.5	343.4	236.3	273.5	338.7	247.1	291.6	341.1
Number	667	667	667	658	658	658	1325	1325	1325
Corr	.13	.17	.17	.04	.06	.07	.10	.10	.10
NMB (%)	-34	-21	-12	-46	-37	-22	-40	-30	-20
NME (%)	52	54	58	59	58	61	60	60	60
NO	Daytime			Nighttime			Daily		
	NOAH	SLAB	TKE	NOAH	SLAB	TKE	NOAH	SLAB	TKE
MeanObs	3.6	3.6	3.6	11.7	11.7	11.7	7.5	7.5	7.5
MeanMod	3.1	3.5	5.2	.02	.02	.08	1.6	1.8	2.8
Number	395	395	395	363	363	363	758	758	758
Corr	.36	.36	.57	.29	.32	.27	-.07	-.07	-.02
NMB (%)	-13	-4	45	-100	-100	-99	-78	-76	-63
NME (%)	79	83	100	100	100	99	94	96	99
NO <sub>2</sub>	Daytime			Nighttime			Daily		
	NOAH	SLAB	TKE	NOAH	SLAB	TKE	NOAH	SLAB	TKE
MeanObs	11.9	11.9	11.9	14.7	14.7	14.7	13.2	13.2	13.2
MeanMod	13.3	16.1	18.0	29.5	32.4	40.4	21.0	23.9	28.7
Number	805	805	805	730	730	730	1535	1535	1535
Corr	.32	.31	.32	.17	.21	.21	.28	.30	.30
NMB (%)	12	36	52	100	120	170	89	80	120
NME (%)	84	98	1.04	130	140	190	110	120	150
PM <sub>2.5</sub>	Daytime			Nighttime			Daily		
	NOAH	SLAB	TKE	NOAH	SLAB	TKE	NOAH	SLAB	TKE
MeanObs	10.3	10.3	10.3	10.4	10.4	10.4	10.3	10.3	10.3
MeanMod	14.1	15.3	16.2	11.1	12.1	12.2	12.7	13.8	14.3
Number	886	886	886	811	811	811	1697	1697	1697
Corr	0.26	0.33	0.37	0.16	0.18	0.2	0.2	0.3	0.3
NMB (%)	36	48	56	7	17	18	20	30	40
NME (%)	66	70	74	51	54	54	60	60	60

For  $\text{SO}_4^{2-}$ , SOR\_TKE simulates values 25-30% higher than SOR\_NOAH or SOR\_SLAB on day 1 at most locations (as seen in Figure 3), while SOR\_SLAB simulates higher maximum concentrations by near 50% at most locations for days 4-5. The simulation results with SOR\_NOAH are similar to those of SOR\_TKE at most locations. Table 2 shows the performance statistics for  $\text{O}_3$ , CO, NO,  $\text{NO}_2$ , and  $\text{PM}_{2.5}$  for daytime and nighttime, as well as daily values. For hourly predictions throughout a day, SOR\_TKE performs better for  $\text{O}_3$  (14.3% vs. 32.3% and 36.0%) and CO (-17.2% vs. -40.0% and -29.2%) mixing ratios in terms of NMB as compared to SOR\_NOAH and SOR\_SLAB. SOR\_NOAH performs better for  $\text{PM}_{2.5}$  than SOR\_SLAB and SOR\_TKE (22.4% vs. 33.3% and 38.0%). While all model simulations reproduce observed daytime  $\text{O}_3$  mixing ratios reasonably well (with NMBs of 5-9%), they all significantly overpredict nighttime  $\text{O}_3$  mixing ratios (with NMBs of 37-94%). This can be partially attributed to the significant underprediction of nighttime NO (as a result of overprediction of mixing at night), which titrates less amounts of  $\text{O}_3$  in the nocturnal PBL. While the CO mixing ratios during some time periods (e.g., 7:00AM to 10:00AM LST on days 1-5) are overpredicted, those during other periods are underpredicted. The underpredictions dominate, resulting in a net negative bias. While NO mixing ratios are significantly underpredicted (by 63-78%),  $\text{NO}_2$  mixing ratios are significantly overpredicted (80-120%), with much larger deviations occurring at night for both species. In addition to inaccuracies in the meteorological predictions, the equilibrium assumption made in simulating nitrogen chemistry in the RADM2 gas-phase mechanism may not represent the ambient conditions simulated for this particular episode, which may contribute to the larger model biases for  $\text{NO}_x$  predictions. An interesting observation is that each model shows a lower NMB during daytime simulations of  $\text{O}_3$ , CO, NO and  $\text{NO}_2$ , but daytime biases for  $\text{PM}_{2.5}$  are higher than those at night. While the secondary PM species may have been well reproduced during daytime, the larger uncertainties may be likely due to the uncertainties in the emissions of primary PM species such as BC and OM. Among the three model simulations, SOR\_TKE gives the best performance for  $\text{O}_3$ , CO, and NO, and SOR\_NOAH gives the best performance for  $\text{NO}_2$  and  $\text{PM}_{2.5}$ . It is noted that differences between observations and model predictions from the three simulations are larger than differences among model results. This indicates that uncertainties other than meteorology

(e.g., emissions) may also be important factors for the discrepancies between observations and predictions.

#### 4. SUMMARY

Our evaluation has shown, for this particular episode, that meteorological schemes play an important role in the simulation of chemical species. It appears that the use of different PBL parameterization and land-surface schemes affects chemical predictions for this scenario. When comparing the performance of the models in terms of meteorological variables and chemical species, the SOR\_SLAB simulation performs better in terms of temporal and statistical analyses than either SOR\_NOAH or SOR\_TKE, with the exception of SOR\_NOAH predicting surface temperature more accurately. Although SOR\_SLAB gives better meteorological predictions, SOR\_TKE performs better than SOR\_NOAH and SOR\_SLAB for  $\text{O}_3$  and CO predictions. A better representation of daytime

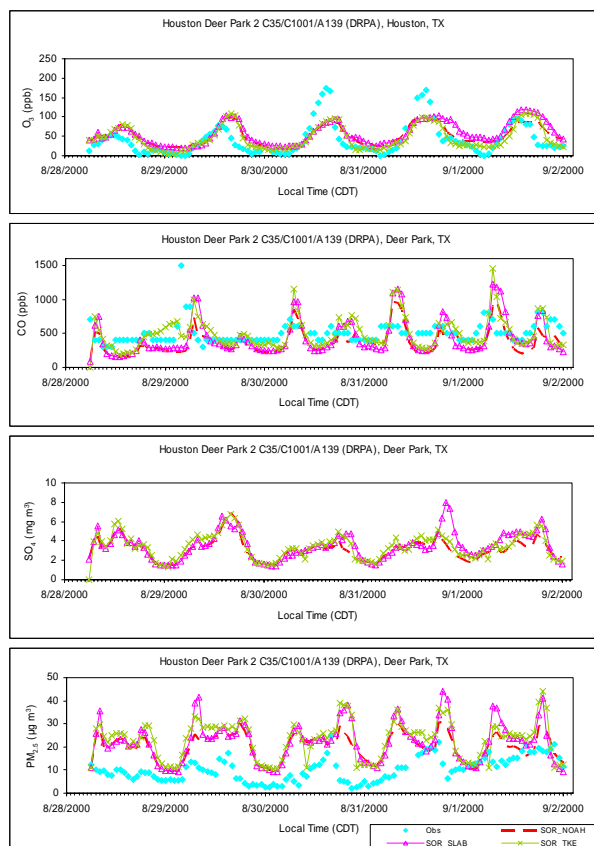


Figure 3. Temporal distributions of (a)  $\text{O}_3$ , (b) CO, (c)  $\text{SO}_4^{2-}$ , and (d)  $\text{PM}_{2.5}$  at Deer Park, Texas, during 28 August - 2 September 2000 (Observed  $\text{SO}_4^{2-}$  data was not available during this time period).

and nocturnal PBL heights by the TKE PBL scheme gives better predictions for primary species such as CO and NO (higher than those predicted by the YSU scheme). Compared to SOR\_NOAH, higher NO mixing ratios predicted by SOR\_TKE titrate more O<sub>3</sub> at night, resulting in a lower NMB in nighttime O<sub>3</sub> predictions (and thus overall O<sub>3</sub> predictions because of the dominance of the nighttime NMB). For NO<sub>2</sub> and PM<sub>2.5</sub>, SOR\_NOAH gives the best overall performance.

A strong correlation has typically been considered between temperature and mixing ratios of O<sub>3</sub>. An interesting finding of this study, however, shows that better temperature predictions do not necessarily give a better O<sub>3</sub> performance. While SOR\_NOAH simulates temperature more accurately than either SOR\_SLAB or SOR\_TKE, its O<sub>3</sub> performance is not as good as that of SOR\_TKE (with a NMB of 32.3% and 14.3%, respectively). This is because the O<sub>3</sub> performance for this episode is dominated by the poor O<sub>3</sub> performance at night, rather than daytime. The poor O<sub>3</sub> performance at night by SOR\_NOAH is due partially to the fact that the YSU scheme gives poorer representation of the nocturnal PBL than the TKE scheme.

Further study is necessary to fully understand the impacts of meteorological variables on the formation and transport of chemical species and all major likely causes for the discrepancies in meteorological and chemical predictions with different meteorological schemes.

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