Bias in evaluating chemical transport models with maximum daily 8-hour average (MDA8) surface ozone for air quality applications

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- 9 Key Points:
- Evaluation of air quality models with the MDA8 ozone metric requires simulation of the ozone diurnal cycle but models are biased at night.
- Simulation of nighttime ozone is challenging due to the day-night transition in atmospheric stability and plant stomata closure.
- Models fail to capture frequent occurrences of MDA8 ozone <20 ppb on rainy days due to missing surface stratification or ozone deposition.
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17 Abstract

Chemical transport models typically compare simulated surface ozone concentrations to 18 observations of the maximum daily 8-hour average (MDA8), the standard air quality policy 19 metric. This requires successful simulation of the surface ozone diurnal cycle including 20 nighttime depletion, but models are generally biased at night. We quantify the problem with the 21 22 GEOS-Chem model for the Southeast US during the 2013 NASA SEAC⁴RS aircraft campaign. The model is unbiased relative to the daytime mixed layer aircraft observations but has a + 5 ppb 23 bias relative to MDA8 surface ozone observations. The model also does not capture observed 24 occurrences of <20 ppb MDA8 ozone on rainy days. Restricting the evaluation to afternoon 25 hours and dry days removes the bias. Better understanding of surface layer stratification and 26 ozone depletion under nighttime and rainy conditions is needed. Resolving the timing in the day-27 night transition in atmospheric stability and its correlation with plant stomata closure is critical. 28

29 **1 Introduction**

30 Ground-level ozone is harmful to human health and vegetation. It is produced when volatile organic compounds (VOCs) and carbon monoxide (CO) are photochemically oxidized in the 31 presence of nitrogen oxide radicals (NO_x \equiv NO+NO₂). Ozone air quality standards in different 32 countries are generally formulated using the maximum daily 8-hour average concentration 33 (MDA8) as a metric. In the US, the current ozone National Ambient Air Quality Standard 34 (NAAQS) is 70 ppb as the fourth-highest MDA8 concentration per year, averaged over three 35 years (EPA, 2015). Exceedances of the standard generally occur during daytime due to 36 37 photochemical production and to entrainment of elevated ozone from aloft (Kleinman, et al., 1994). Ozone is depleted at night due to deposition and chemical loss in a shallow surface layer 38 capped by a stratified atmosphere. 39

40 Air quality agencies rely on chemical transport models (CTMs) to identify the most effective emission reduction strategies for ozone pollution. CTMs predict surface ozone concentrations on 41 the basis of NO_x, VOC, and CO emissions, accounting for chemistry and meteorological 42 conditions. MDA8 ozone is commonly used as the metric for evaluating models with 43 observations and making predictions relevant to air quality standards (Fiore et al., 2009; Mueller 44 and Mallard, 2011; Emery et al., 2012; Lin et al., 2012; Rieder et al., 2015). Use of this metric 45 implicitly requires successful simulation of the diurnal cycle in surface ozone but models are 46 47 generally too high at night, apparently because they cannot resolve the local stratification and associated depletion from surface deposition. This is a problem not only in global models with 48 coarse vertical resolution (Lin and McElroy, 2010; Schnell et al., 2015; Strode et al., 2015) but 49 also in regional air quality models (Herwehe et al., 2011; Solazzo et al., 2012). 50

51 Here we evaluate the use of the MDA8 ozone metric in the GEOS-Chem CTM, a global model frequently used in studies of regional ozone air quality and evaluated for this purpose with 52 MDA8 ozone (Racherla and Adams, 2008; Lam et al., 2011; Zhang et al., 2011; Zoogman et al., 53 2011; Emery et al., 2012; Zhang et al., 2014). In our previous application of the model to the 54 Southeast US during the NASA SEAC⁴RS aircraft campaign in August-September 2013 (Travis 55 et al., 2016), we found that the model had no significant bias relative to aircraft ozone 56 57 observations below 1 km altitude but overestimated MDA8 surface ozone by +6 ppb on average. As we show here, this may largely be explained by the poor representation of surface layer 58

59 stratification. The ultimate solution of this problem will require improved representation of 60 boundary layer physics, but we propose in the meantime some simple corrective measures.

61 2 Comparing simulations of mixed layer and MDA8 surface ozone

The GEOS-Chem simulation used here was previously applied by Travis et al. (2016) to interpret 62 observations from the SEAC⁴RS aircraft campaign in August-September 2013 (Toon et al., 63 2016). It is based on GEOS-Chem version 9.02 with detailed oxidant-aerosol chemistry 64 65 (www.geos-chem.org) and is driven by assimilated meteorological data from the Goddard Earth Observing System - Forward Processing (GEOS-FP) product of the NASA Global Modeling and 66 Assimilation Office (GMAO) using the GEOS-5.11.0 general circulation model (Molod et al., 67 2012). The GEOS-FP data have a native horizontal resolution of 0.25° latitude by 0.3125° 68 longitude, with 72 levels in the vertical on a hybrid sigma-pressure grid and a temporal 69 resolution of one hour for surface variables and mixing depths. This native resolution is used in 70 71 GEOS-Chem over North America and adjacent oceans (130° - 60° W, 9.75° - 60° N), with boundary conditions from a global simulation with $4^{\circ} \times 5^{\circ}$ horizontal resolution. The lowest levels 72 are centered at about 65 m, 130 m, 200 m, and 270 m above ground level (AGL). Boundary layer 73 74 turbulence follows the clear-sky non-local parameterization from (Holtslag & Boville, 1993), as implemented in GEOS-Chem by (J.-T. Lin & McElroy, 2010). Detailed evaluations of GEOS-75 Chem with observations over the Southeast US for the SEAC⁴RS period are presented in other 76 papers (Kim et al., 2015; Marais et al., 2016; Yu et al., 2016; Zhu et al., 2016; Miller et al., 77 2017;). Specific evaluation for ozone is presented in Travis et al. (2016). 78

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Travis et al. (2016) found that despite successful simulation of ozone observations from the 80 SEAC⁴RS aircraft in the mixed layer below 1 km altitude, MDA8 surface ozone was biased by 81 +6 ppb on average. Fig. 1 (left panel) shows the probability density functions (pdfs) of ozone 82 concentrations measured by the aircraft (12-17 local solar time or LT) and simulated by the 83 84 model along the flight tracks. Model values are adjusted to local solar time by 1 hour per 15° longitude. The data have been filtered for biomass burning (CH₃CN > 200 ppt) and urban plumes 85 $(NO_2 > 4 \text{ ppb})$. The bias between the model and observations is small (+2 ppb) and not 86 87 statistically significant (p=0.07). The center panel of Fig. 1 shows the observed and simulated pdfs of daily MDA8 surface ozone in August-September 2013 at the thirteen rural CASTNET 88 sites in the Southeast US (EPA, 2018). The model is biased high by +8 ppb on average and this 89 90 is highly significant (p < 0.01). The bias differs slightly from the +6 ppb in Travis et al. (2016) who showed a comparison for June-August. Comparison of the mean aircraft and MDA8 surface 91 concentrations in Figure 1 indicates a vertical difference of 9 ppb in the observations but only 3 92 93 ppb in GEOS-Chem.

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95 **3 Correcting for surface layer gradients**

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A first problem in comparing the model to the CASTNET surface air observations is the mismatch between the lowest model level midpoint ($z_m = 65$ m above ground) and the level at which the observations are made ($z_1 = 10$ m). The model in fact implicitly simulates an ozone concentration at z_1 through the aerodynamic resistance $R_a(z_1, z_m)$ to turbulent vertical transfer in the resistance-in-series parameterization of dry deposition (Brasseur & Jacob, 2017). The model calculates a local ozone deposition velocity $v_d(z_m)$ at altitude z_m assuming uniformity of the 103 vertical flux down to the surface. We can then infer the implicit model ozone concentration $C(z_1)$ 104 at 10 m from the explicit concentration $C(z_m)$ at 65 m (Zhang et al., 2012):

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$$C(z_1) = (1 - R_a(z_1, z_m)v_d(z_m))C(z_m)$$
 (1)

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107 $R_a(z_l, z_m)$ is calculated in GEOS-Chem by similarity with momentum for a neutral atmosphere 108 (friction velocity u^*) with a heat-based stability correction $\phi_h(z/L)$ where *L* is the Monin-

109 Obukhov length and *k* is the von Karman constant:

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111
$$R_a = \int_{z_1}^{z_m} \frac{\phi_h(z/L)}{ku^* z} dz$$
 (2)

Equations 3(a-c) describe ϕ_h , from Dyer (1974) for unstable and moderately stable conditions (*z/L* < 1) and from Holtslag et al. (1990) for stable conditions (*z/L* > 1):

114 $\phi_h = 5 + z/L,$ z/L > 1 3(a)

115
$$\phi_h = 1 + 5 z/L$$
, $0 < z/L < 1$ 3(b)

116
$$\phi_h = (1 - 16 z/L)^{-1/2}, \quad z/L < 0$$
 3(c)

The model deposition velocity $v_d(z_m)$ over the Southeast US during SEAC⁴RS averages 0.7 \pm 0.3 117 cm s⁻¹ in daytime, consistent with observations (Travis et al., 2016). Applying the correction 118 from equation (1) at the CASTNET sites we find a mean MDA8 model concentration at 10 m 119 altitude of 45 ± 8 ppb, as compared to 48 ± 9 ppb at 65 m. Correcting the model to 10 m altitude 120 121 thus decreases the model bias relative to observations by 3 ppb, but a bias of +5 ppb remains. Model MDA8 ozone at 65 m has ten exceedances of the 70 ppb NAAOS for the CASTNET data 122 in Figure 1, as compared to one in the observations, and sampling the model at 10 m decreases 123 124 the number of exceedances to four.

125 **4 Segregating rainy conditions**

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The most severe bias in comparing the model MDA8 ozone to the CASTNET observations in Figure 1 is for the low tail of the distribution (less than 25 ppb). 7 % of observed MDA8 ozone values are below 25 ppb but there is only one value below 25 ppb in the model at either 65 or 10 m. This low-tail bias cannot be simply explained by inflow of low-ozone tropical air from the Gulf of Mexico (Fiore et al., 2002; McDonald-Buller et al., 2011) because the model simulation is unbiased over the Gulf of Mexico relative to the SEAC⁴RS aircraft observations (Travis et al., 2016).

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We find instead that the low MDA8 ozone values in the CASTNET observations are associated 135 with rainy conditions and that rain has less effect on ozone in the model. Figure 2 segregates the 136 frequency distribution of MDA8 ozone at CASTNET sites between rainy days and dry days. 137 Observed ozone on rainy days averages 9 ppb lower than on dry days (33 vs 42 ppb). Model 138 ozone is also lower on rainy days but not by as much (41 vs 46 ppb). Rainy conditions can cause 139 MDA8 ozone to drop below 20 ppb in the observations but not in the model. Depletion of 140 surface ozone under rainy conditions is not due to wet scavenging, considering the low solubility 141 of ozone in water, but likely reflects vertical stratification from surface evaporative cooling. 142

Rainfall or dew may also enhance the non-stomatal component of ozone dry deposition (Finkelstein et al., 2000; Altimir & Kolari, 2006; Potier et al., 2017) but the mechanism for this enhancement is uncertain. Comparing the 10-m model MDA8 concentration to observations excluding rainy days decreases the model mean bias modestly from +5 ppb to +4 ppb, but more importantly it excludes the low tail of the observed distribution that the model cannot capture.

148 **5 Accounting for diurnal bias**

Yet another factor in the model overestimate of MDA8 surface ozone is the poor simulation of 149 the diurnal cycle. Figure 3 shows the average ozone diurnal cycle for dry days in the model and 150 observations at the CASTNET sites from Fig. 1. The observations show a typical diurnal cycle of 151 maximum values in early afternoon (14-16 LT) and gradual decrease at night to a mean 152 minimum value of 17 ppb at 7 LT. The nighttime depletion cannot be due to chemical titration 153 by anthropogenic NO emissions since the CASTNET sites are rural and must instead be due to 154 deposition and/or titration by short-lived biogenic VOCs (Goldstein et al., 2004; Ruuskanen et 155 al., 2011; Rossabi et al., 2018) under stratified surface layer conditions. The model diurnal cycle 156 157 at 65 m altitude (lowest model level) has the correct phase but the amplitude is much too weak. Correcting the model to 10 m altitude increases the amplitude but nighttime depletion is still too 158 weak. The difference between 65 and 10 m grows rapidly between 16 and 18 LT as the 159 atmosphere becomes stable (L > 0) and the mixed layer collapses but ozone deposition is still fast 160 because of open stomata. After the stomata close at night the gradient weakens. We find 161 negligible difference in the model diurnal cycle shown in Figure 3 between August and 162 163 September. Silva & Heald (2018) show that the low nighttime ozone deposition velocities in the model are consistent with observations, which would include the effect of titration by nighttime 164 emissions of short-lived biogenic VOCs. Lack of diurnal cycle in modeled anthropogenic 165 emissions has been suggested as a cause of the general underestimate among models of the 166 summertime diurnal amplitude of ozone concentrations (Schnell et al., 2015), but the emissions 167 used here have hourly resolution based on the National Emission Inventory of the US 168 Environmental Protection Agency. We conclude that the insufficient nighttime depletion in the 169 model must be due to insufficient vertical stratification of the surface layer, combined with poor 170 resolution of the correlated timing between day-night transition to stable conditions and stomata 171 closure. 172

A consequence of the insufficient model depletion of ozone at night is that the model may err in 173 the diurnal timing of MDA8 ozone. Fig. 4 shows the pdf of the beginning of the 8-hour interval 174 for MDA8 ozone at the CASTNET sites on dry days, comparing the observations and the model. 175 In the observations the pdf peaks sharply at 11 LT (MDA8 window of 11-18 LT), consistent with 176 the mean diurnal cycle of Figure 3. The model sampled at 65 m also has a maximum probability 177 of MDA8 ozone starting at 11 LT, but also a secondary maximum at 19 LT that is absent from 178 the observations. The latter conditions occur in the model when the atmosphere becomes stable 179 already at 16 LT, decoupling 65 m from the surface and the associated deposition. Under these 180 conditions the model concentration at 65 m remains high in the evening and at night. Correcting 181 the model calculation of MDA8 to use the 10-m ozone largely removes this secondary maximum 182 (Figure 4) but shifts the peak occurrence of MDA8 ahead by two hours (starting at 9 LT) because 183 184 of the exaggerated model drop at 17 LT when the model atmosphere becomes stable but ozone stomatal deposition is still active (Fig. 3). The transition from a convective mixed layer to stable 185 nighttime conditions is difficult for models to capture and is an active area of research (Lothon et 186

al., 2014). The correlated timing with stomatal closure further complicates the simulation of theday-night transition in surface ozone.

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Model error in the simulation of the ozone diurnal cycle due to insufficient nighttime depletion 190 thus induces a representation error when comparing to MDA8 observations, as the MDA8 191 periods in the model do not correspond to the same times of day as in the observations. This 192 causes positive bias in the comparison. From the standpoint of evaluating the broader processes 193 controlling ozone in the model, the nighttime bias may be of little importance as it only affects a 194 shallow surface layer. In that case a solution is to focus on afternoon conditions for model 195 evaluation, as is done in many studies (i.e. Fiore et al., 2002). The right panel of Figure 1 196 compares simulated and observed pdfs of surface ozone at the CASTNET sites at 12-17 LT on 197 dry days, sampling the model at 10 m altitude. The +8 ppb bias in the original model comparison 198 (center panel) is reduced to an insignificant +1 ppb. 199

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201 6 Implications

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203 We identified three modeling problems biasing the comparison to observed MDA8 ozone for air quality applications: (1) vertical mismatch between the lowest model level and the altitude of the 204 observations, (2) insufficient vertical stratification and/or ozone loss under rainy conditions, and 205 206 (3) inadequate representation of the day-night transition to stable conditions leading to error in timing of the 8-hour MDA8 window. Problem (1) is readily solved by using the parameterization 207 of surface layer turbulence implicit in the model simulation of dry deposition. Problems (2) and 208 (3) suggest the need for more research in the dynamics of stable boundary layers but can be 209 circumvented by focusing model comparisons to observations on dry conditions and afternoon 210 hours. 211

212 Data Availability

PRISM and precipitation be downloaded temperature data can 213 at http://www.prism.oregonstate.edu/historical/. CASTNET observations are available here: 214 https://www.epa.gov/castnet. SEAC⁴RS aircraft observations are available here: https://www-215 air.larc.nasa.gov/cgi-bin/ArcView/seac4rs. 216

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351 352 Figure 1 - Probability density functions (pdfs) of ozone concentrations in the Southeast US (94.5-80 W, 29.5-38 N) in August-353 September 2013, sampled at the blue locations in the maps inset. Observations are compared to GEOS-Chem model values 354 sampled at the same locations and times. Means and standard deviations are given inset for each pdf. The left panel shows 355 afternoon (12-17 local solar time) mixed layer values from the SEAC⁴RS DC8 aircraft at 0.4-1.0 km altitude. Ozone 356 measurements are from the NOAA NOvO3 four-channel chemiluminescence (CL) instrument (Rverson et al., 1998) The center 357 panel shows MDA8 surface ozone at the CASTNET network of 13 rural sites, compared to the model sampled at 65 m (dashed 358 line) above ground (lowest model gridpoint) and the inferred model value at 10 m (solid line) as described in the text. The right 359 panel shows afternoon ozone at the CASTNET sites excluding days with rain in either the model or the observations.

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Figure 2 - Probability density functions (pdfs) of MDA8 ozone at CASTNET sites in the Southeast US in August-September 363 2013, segregating rainy and dry days. Here we define rainy days in both the observations and the model by 24-h total rainfall

exceeding 6 mm and dry days by 24-h total rainfall less than 1 mm. Observed rainy and dry days are diagnosed with observations

from the PRISM climate group (PRISM, 2016) regridded to the model resolution of $0.25^{\circ} \times 0.3125^{\circ}$. Model rainy and dry days are diagnosed from the GEOS-FP data. The model is sampled at 10 m altitude to match observations, as described in Section 3.

For each sky condition, the mean ozone and its standard deviation are given inset with the frequency of that sky condition in

368 parentheses. The probabilities of dry and rainy condition do not add to 100 % because we do not include marginal days where

rainfall is between 1 and 6 mm.

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Figure 3 – Mean diurnal cycle of ozone and related surface variables at the 13 Southeast US CASTNET sites in Figure 1 for August-September 2013. Ozone observations in the top left panel are compared to GEOS-Chem values sampled at 65 m altitude (lowest model level) and at 10 m altitude (where the observations are sampled). Other panels show the mean 10-m ozone deposition velocity in GEOS-Chem, the median Monin-Obukhov length L in the GEOS-FP data used to drive GEOS-Chem, and the mean mixed layer depth in the GEOS-FP data. Days where precipitation exceeds 1 mm in either the model or observations are excluded. Local hour refers to solar time (maximum solar elevation at noon). Vertical dashed lines at 6, 12, and 18 local time are to guide the eye.



Beginning hour of MDA8 ozone interval

Figure 4 – Timing of MDA8 ozone at the Southeast US CASTNET sites in August-September 2013. The figure shows the probability density functions (pdfs) of the beginning hour of the 8-hour period defining the MDA8 ozone value for each day.
Only dry days (24-h precipitation less than 1 mm) are included.

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