High resolution CMAQ simulations of ozone exceedance events during the Lake Michigan Ozone Study

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18 Abstract: We evaluate two high-resolution Lake Michigan air quality simulations during the 2017 Lake Michigan 19 Ozone Study campaign. These air quality simulations employ identical chemical configurations but use different input 20 meteorology. The "EPAAP-XM-obsAP-XM" configuration follows EPA recommended modeling practices, whereas 21 the "YNT_SSNG" employs different parameterization schemes and satellite-based inputs of sea surface temperatures, 22 green vegetative fraction, and soil moisture and temperature. Overall, we find similar performance in model 23 24 simulations of hourly and daily 8-hour maximum (MDA8) ozone, with the <u>AP-XM-obsAP-XM</u>EPA and YNT_SSNG simulations showing biases of -113.4231 and -13.54 ppbv, respectively during periods when the observed MDA8 was 25 26 greater than 70ppbv. However, for the two monitoring sites that observed high ozone events, the <u>AP-XM_obsAP-</u> XMEPA simulation better matched observations at Chiwaukee Prairie Sheboygan KA and the YNT_SSNG simulation 27 better matched observations at -Sheboygan KAChiwaukee Prairie. We find differences between the two simulations 28 29 are largest for column amounts of ozone precursors, particularly NO2. Across three high ozone events, the YNT_SSNG simulation has a lower column NO₂ bias (0.17 x 10¹⁵ molecules/cm²) compared to the AP-XM-obsAP-XMEPA 30 simulation (0.3135 x 1015 molecules/cm2). The YNT_SSNG simulation also has an advantage in better capturing the 31 structure of the boundary layer and lake breeze during the June 2 high ozone event, although the timing of the lake 32 breeze is about 3 hours too early at Sheboygan. Our results are useful in informing an air quality modeling framework 33 for the Lake Michigan area.

34 1. Introduction

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35 Ground-level ozone has many well-documented effects on human health, including increased risk for respiratory and 36 cardiovascular diseases, and even premature death (Di et al., 2017; Lelieveld et al., 2015; Manisalidis et al., 2020). 37 38 Ozone also damages plant tissue, affecting crop health (e.g. Clifton et al., 2020; Shindell et al., 2012). Ground -level ozone is formed by photochemical reactions between nitrogen oxides (NOx) and volatile organic compounds (VOCs); 39 major NOx sources include fuel combustion, biomass burning, soil microbes, and lightning, with anthropogenic 40 sources dominant (Hall et al, 1996; Juncosa Calahorrano et al., 2021; Lamsal et al., 2010; Lawrence and Crutzen, 41 1999; Nault et al., 2017), major sources of VOCs include industrial processes and natural sources, such as trees 42 (Guenther et al., 1995; He et al., 2019). 43

Since the first National Ambient Air Quality Standard (NAAQS) for ozone was released in 1979, most lakeshore counties in the states bordering Lake Michigan (Wisconsin, Illinois, Indiana, and Michigan) have been designated as being in nonattainment for surface ozone in one or more of the subsequent NAAQS revisions. These states are required by the Clean Air Act to develop State Implementation Plans (SIPs) to demonstrate strategies to bring affected areas into attainment and to mitigate the impacts of high ozone concentrations. Large decreases in local emissions of ozone

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49 precursors have steadily reduced one- and eight-hour maximum ozone concentrations across the region in recent 50 decades (Adelman 2020). However, the implementation of stricter ozone NAAQS, along with increases in susceptible populations (e.g. Daggett et al., 2000), means that additional air quality modeling assessments are necessary to help states demonstrate that they can reach attainment by the required statutory deadlines.

51 52 53 54 55 The areas along the Lake Michigan shoreline are susceptible to high ozone amounts because of a combination of abundant precursor emissions and transport processes, particularly the lake breeze circulation. The relationships 56 between area emissions and meteorology as they impact air quality along the Lake Michigan shoreline have been 57 characterized in field campaigns (Sexton and Westberg, 1980; Dye et al., 1995; Foley et al., 2011; Stanier et al., 2021), 58 and the meteorological component is the subject of Part 1 of this study (Otkin et al., 20232). Ozone concentrations 59 60 along coastlines can be enhanced significantly when urban emissions react within the shallow, stable, marine boundary layer (Fast and Heilman, 2003). The lake breeze circulation is particularly important for enhanced ozone production 61 over Lake Michigan where it contributes to roughly 80% of high ozone episodes observed in eastern Wisconsin 62 (Lennartson and Schwartz, 2002; Cleary et al., 2021). Lake breeze circulations impact ozone concentrations elsewhere 63 in the Great Lakes including southern Ontario, Michigan, and Ohio (Makar et al, 2010, He et al., 2011, Brook et al, 64 2013, Stroud et al, 2020).

66 As highlighted by Dye et al. (1995), there has been a need for a modeling framework that represents the finer scales 67 of emissions transport and chemistry near the Lake Michigan shoreline. It is our opinion that dDeveloping emission control strategies to mitigate these coastal high ozone events requires accurate prediction of the lake breeze transport 68 69 processes at scales of 1-10 km. Furthermore, tThese chemical transport processes cannot be accurately resolved using 70 71 the 12-km resolution meteorological and chemical simulations typically used in air quality modeling for previous SIPs

72 73 74 We have developed a high-resolution, satellite-constrained meteorological modeling platform for the Midwest United States that supports the needs of the Lake Michigan Air Directors Consortium (LADCO) as they conduct detailed air 75 76 77 quality modeling assessments for its member states. In part I of this study, Otkin et al. (20232) assessed the impact of different high-resolution surface datasets, parameterization schemes, and analysis nudging on near-surface meteorological conditions and energy fluxes relative to the model configuration and input datasets typically employed 78 79 by the Environmental Protection Agency (EPA). In part II of this study, we use meteorological output obtained from two of these simulations, as input to the EPA Community Multiscale Air Quality (CMAQ) model version 5.2.1 (Byun 80 and Schere, 2006; Nolte et al., 2015) model simulations to assess the impact of these model changes on ozone forecasts 81 in the Lake Michigan region. The remainder of this paper is organized as follows:- Section 2 contains a description of 82 the CMAQ model configurations and observational data used for evaluation;- rResults are presented in Section 3, with 83 discussion and conclusions provided in Section 4.

84 2. Methods

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85 In this work, we compare two CMAQ simulations, one with AP-XM-obsbaseline EPA meteorology, and the other 86 with meteorology from our optimized WRF configuration, as es-detailed in Part I (Otkin et al., 20232). Both sets of 87 meteorological simulations employ a triple-nested domain configuration containing 12-, 4-, and 1.33333 (1.3)-km 88 resolution grids, respectively (Figure 1 in Otkin et al., 20232), constrained to 6-hourly, 0.25-degree GFS Final 89 reanalyses and as well as using RRTMG longwave and shortwave radiation (Iacono et al. 2008; Mlawer et al. 1997) 90 on all three domains, and the Kain-Fritsch cumulus scheme (Kain 2004) on the outer two domains, and explicit 91 convection on the innermost domain. Both simulations have the same vertical resolution throughout, with 6 model 92 layers below 200m, 4 model layers below 100m, and the lowest three layers at ~9, 27, and 55m above ground level. 93 The AP-XM obsAP-XMEPA simulation employs the Morrison microphysics (Morrison et al. 2005), ACM2 PBL 94 (Pleim 2007), and the Pleim-Xu LSM (Gilliam and Pleim 2010; Xiu and Pleim, 2001) parameterization schemes, 95 which are the same schemes used within CMAQ and is therefore considered our baseline meteorological simulation, 96 with indirect soil moisture and soil temperature nudging and nudging to surface observations of humidity, sea level 97 pressure, temperature and wind from the Meteorological Assimilation Data Ingest System (MADIS, 98 https://madis.ncep.noaa.gov/). Our optimized meteorological modeling platform uses the YSU PBL (Hong et al. 99 2006), Noah LSM (Chen and Dudhia, 2001; Ek et al. 2003), and Thompson microphysics (Thompson et al. 2008, 100 2016) schemes, constrained by high-resolution (1km) soil moisture and temperature analyses (Case 2016; Case and 101 Zavodsky 2018; Blankenship et al. 2018) from the Short-term Prediction Research and Transition Center (SPoRT),

102 daily high resolution (1.3 km) Great Lakes surface temperatures (Schwab 1992) from the Great Lakes Surface 103 Environmental Analysis (GLSEA), and high resolution (4 km) Green Vegetation Fraction (GVF) from the Visible 104 Infrared Imaging Radiometer Suite (VIIRS; Vargas et al. 2015) in place of monthly GVF climatologies. This 105 optimized configuration is hereafter referred to as the YNT_SSNG. Otkin et al. (20232) found that the AP-106 XM_obsAP-XMEPA configuration generally produced more accurate meteorological analyses on the 12-km domain, 107 but its accuracy decreased with finer model grid resolution. In contrast, the YNT_SSNG statistics showed consistent 108 reductions in root-mean-square error (RMSE) for 2-m temperature, 2-m water vapor mixing ratio, and 10-m wind 109 speed relative to the AP-XM_obsAP-XMEPA as the model resolution increased from 12 km to 1.3 km. We note that 110 dDifferences in near surface wind speed and GVF will also impact deposition velocities in the CMAQ simulations. 111

112 Each CMAQ simulation is run with the same configuration and anthropogenic emissions. Using CMAQv5.2.1 113 (Appel et al., 2017; US EPA, 2018), Que configuration includes "AERO6" aerosol chemistry, the Carbon Bond 6 114 chemical mechanism revision 3 (CB6r3; Emery et al., 2015; Luecken et al., 2019), and in-line photolysis. CMAQ 115 was run with 39 vertical layers with a top of approximately 100 hPa, thus using all available layers from our WRF 116 simulations. As with our WRF simulations, we ran CMAQ on three domains: one using 12 km by 12 km horizontal 117 resolution over the continental U.S. (396 x 246 grid points), one using 4 km by 4 km horizontal resolution over the 118 upper Midwest (447 x 423 grid points), and one using 1.3 km by 1.3 km horizontal resolution over Lake Michigan 119 and nearby areas (245 x 506 grid points). The 12 km CMAQ simulations employ lateral boundary conditions (LBC) 120 from the global Real-time Air Quality Modeling System (RAQMS) model (Pierce et al., 2007), which includes 121 assimilation of ozone retrievals from the Microwave Limb Sounder (MLS) and Ozone Monitoring Instrument (OMI) 122 on the NASA Aura satellite and assimilation of aerosol optical depth (AOD) from the Moderate Resolution Imaging 123 Spectroradiometer (MODIS) on the NASA Terra and Aqua satellites. Utilizing RAQMS LBC for CMAQ 124 continental scale simulations has been shown to significantly increase upper tropospheric ozone and improve daily 125 maximum surface O₃ concentrations (Song et al, 2008) and improve agreement with OMI tropospheric ozone 126 column (Lee et al, 2012) relative to fixed LBC. The 4-km and 1.3-km simulations employ lateral boundary 127 conditions from the respective parent grid.

128 129 Anthropogenic emissions for the 12 km domain were taken from the 2016 National Emissions Inventory Collaborative 130 (NEIC, 2019), version 1. Anthropogenic emissions for the 4 km and 1.3 km domains were taken from the 2017 131 National Emissions Inventory, version 1 (US EPA, 2021; Adams, 2020), where emissions on the 4 km domain were 132 provided by the EPA (Kirk Baker, personal communication), and then interpolated and downscaled -by 1/9th for use 133 on the 1.3 km domain. We acknowledge that the use of downscaled 4km emissions will degrade the performance of 134 135 the 1.3km simulations, but generating 1.3km area emissions from the Sparse Matrix Operator Kernel Emissions (SMOKE) programs was beyond the scope of this project. Biogenic emissions were calculated in-line_using the 136 Biogenic Emission Inventory System (BEIS) with the Biogenic Emissions Landuse Database, version 3 (BELD3; 137 138 Carlton and Baker, 2011). Meteorologically-sensitive input for in-line-biogenic emissions calculations (such as frost dates) were generated separately for each set of CMAQ simulations using Sparse Matrix Operator Kernel Emir 139 (SMOKE) programs. As biogenic emissions are calculated in-line, they vary among our configurations with differing 140 input meteorology and GVF. 141

142 We focus on the innermost, 1.3 km domain surrounding Lake Michigan, during the 2017 Lake Michigan Ozone Study 143 (LMOS) field campaign (Stanier et al., 2021) which occurred from 22 May -22 June 2017. Our chemical evaluation 144 focuses on ozone and threetwo of its precursors, nitrogen dioxide, and formaldehyde and isoprene, in the surface layer 145 and in the atmospheric column. We employ ozone observations from the Air Quality System (AQS) monitoring 146 network, using the Atmospheric Model Evaluation Tool (AMET) developed by the EPA. We also utilize nitrogen 147 dioxide (NO₂) and formaldehyde (HCHO) in situ observations from an EPA trailer that was deployed at Sheboygan, 148 WI, and NO2 and isoprene measurements from the LMOS Zion supersite (Stanier et al, 2021). Iin situ O3 and wind 149 observations at select monitors that were submitted to the LMOS data repository 150 (https://asdc.larc.nasa.gov/soot/power-user/LMOS/2017). For column evaluation, we employ observations of column 151 NO2 and HCHO from the Geostationary Trace Gas Aerosol Sensor Optimization (GeoTASO; Leitch et al., 2014) 152 instrument taken during LMOS (Judd et al., 2019).

153 3. Results

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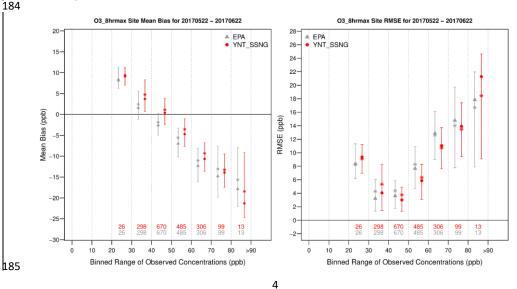
Formatted: Font: (Default) Times New Roman Formatted: Font: (Default) Times New Roman, Subscript Formatted: Font: (Default) Times New Roman Formatted: Font: (Default) AdvTT5843c571, Font color: 154 Our model comparison is organized by three time periods. We first evaluate model performance of the AP-155 XM_obsAP-XMEPA and YNT_SSNG simulations over the entire LMOS period, based on the ozone precursors of 156 NO22_-and-HCHO, and isoprene as well as daily 8-h maximum ozone. Used in the NAAQS for ozone, maximum 8hour ozone amounts are calculated as a rolling 8-h average for each day, starting for the period of 7 am to 3 pm local 157 158 standard time (LST), and ending with the period of 11 pm to 7 am LST the following day. However, 8-h maximum 159 ozone is strongly influenced by days with low and moderate ozone concentrations. Though only 5.9% (112) of the 8-160 hour maximum ozone periods within the 1.3 km domain were above the NAAQS threshold for ozone (70 ppbv) during 161 LMOS (see Fig. -1), it is these higher 8-h maximum ozone values that are most relevant to SIP modeling. 162

163 To evaluate the simulations more precisely, we then evaluate the high ozone days as identified by the two coastal AQS 164 monitors that tend to show the highest ozone concentrations. High ozone days with extensive observations during 165 LMOS 2017 include: 2-4 June, 9-12 June, and 14-16 June (Abdi-Oskouei et al., 2020) and are referred to as events 166 A, B, and C, respectively. Finally, we evaluate model performance over the broader western Lake Michigan shoreline 167 area during the only ozone exceedance event on 2 June.

168 3.1 Model performance over the entire LMOS period

169 3.1.1 8-hour maximum ozone

170 Figure 1 shows binned whisker plots of 8-h maximum ozone bias and RMSE at 10 ppb intervals for the 1.3km 171 simulations for all sites within the 1.3km domain. Systematic high biases for lower ozone concentrations (<~40 ppbv) 172 and a low bias for higher ozone concentrations (> 50 ppbv) are evident in both simulations. THowever, the 173 YNT_SSNG and AP-XM_simulations show similar smaller-biases and RMSE than the AP-XM_obsEPA-for 8-h 174 maximum ozone concentrations between 40-870 ppbya, but Both simulations show similar biases and RMSE within 175 the 70-80 ppby bin and the AP-XM_obsAP-XMEPA shows significantly lower biases and RMSE in the 80-90 ppby 176 bin. Figure 2 shows the geographical distribution of 8-h maximum ozone bias and RMSE for the 1.3km AP-177 XM_obsAP-XMEPA and YNT_SSNG for all AQS sites within the 1.3km domain. Overall biases are largely negative, 178 reflecting underestimates of 8-h maximum ozone at the AQS sites. When compared on a site-by-site basis, the biases 179 and RMSE in 8-h maximum ozone are generally smaller by more than 2 ppbv in the YNT_SSNG simulation with the 180 exception of two AQS sites in North Chicago were the YNT_SSNG simulations shows overestimates of 4-8 ppbv in 181 8-h maximum ozone. -This may be due to the use a more realistic, and lower (relative to climatology) Green 182 Vegetation Fraction (see Figure 2 in Otkin et al, 2023) in the YNT_SSNG simulation which would tend to reduce 183 ozone deposition velocities and increase ozone concentrations (Ran et al, 2016).



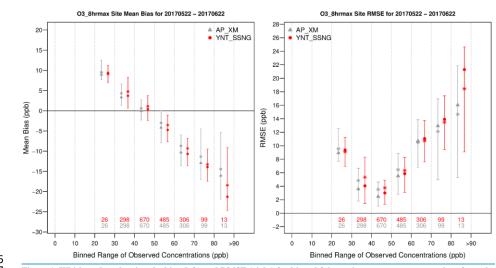
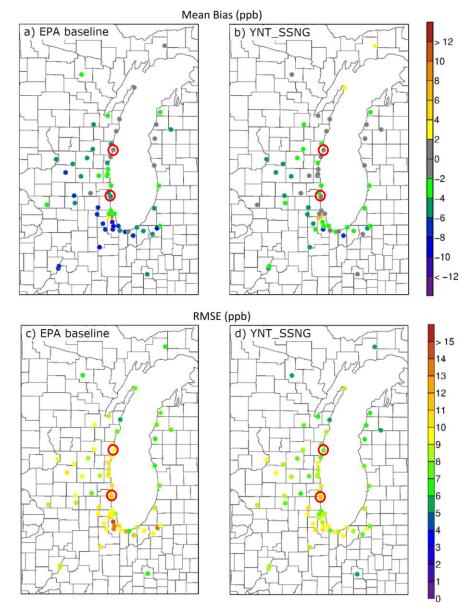




Figure 1. Whisker plots showing the bias (left) and RMSE (right) for binned 8-h maximum ozone concentrations from the <u>AP-XM-obsAP-XMEPA</u> (gray) and YNT_SSNG (red) CMAQ simulations using hourly data within the 1.3km domain during the LMOS period of record from 22 May 2017 to 22 June 2017. Triangles and circles represent the conditional distribution medians, stars represent distribution means, and lines and whiskers represent the Q1 to Q3 ranges.



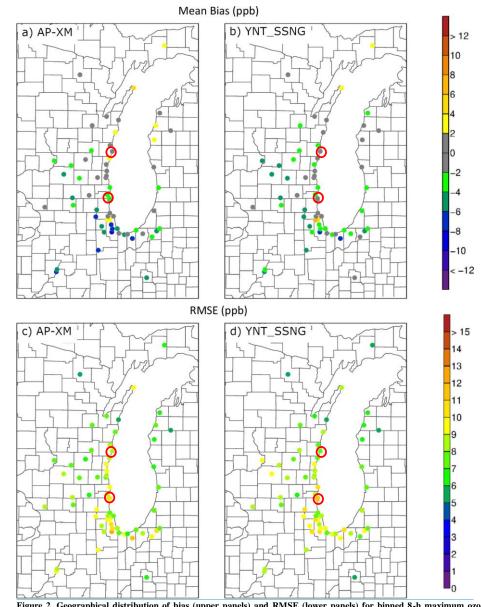


Figure 2. Geographical distribution of bias (upper panels) and RMSE (lower panels) for binned 8-h maximum ozone concentrations from the <u>AP-XM-obsAP-XMEPA</u> (left) and YNT_SSNG (right) 1.3km CMAQ simulations using hourly data from all stations in the 1.3 km resolution inner during the LMOS period of record from 22 May 2017 to 22 June 2017. Bias and RMSE (ppbv) at each site are indicated by the color bar. Two AQS monitors, Sheboygan KA to the north and Chiwaukee Prairie along the Wisconsin-Illinois border are indicated by the red circles.

200 3.1.2 Evaluation with Sheboygan WI ground-based NO2 and HCHO measurements

201 During LMOS, the EPA deployed instruments measuring in-situ NO2 and HCHO in Sheboygan, WI to characterize 202 ozone precursors along the shore of Lake Michigan. These 1-min measurements were taken at Spaceport Sheboygan, 203 which is approximately 9 km north of the Sheboygan, KA monitor highlighted in Figure 2. Here, we use the hourly 204 averaged EPA NO2 and HCHO measurements to evaluate the accuracy of prediction of ozone precursors at Sheboygan 205 for the YNT_SSNG and <u>AP_XM_obsAP-XMEPA</u> CMAQ simulations.

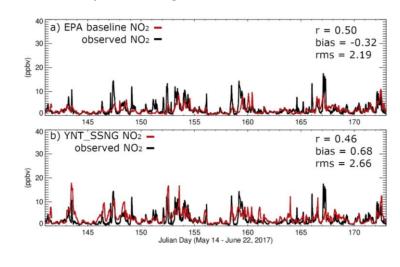
207 Figures 3 and 4 show the hourly NO2 and HCHO comparisons, respectively. There are several periods where observed 208 NO2 (black lines, Figure 3) is above 10 ppby; these periods are generally underestimated by the AP XM_obsAP-209 XMEPA simulation and overestimated by the YNT_SSNG simulation (red lines, Figure 3). We find an overall slight 210 positivenegative bias of -0.1932 ppbv for the <u>AP-XM_obsAP-XMEPA</u> and an overall positive bias of 0.68 pbbv for 211 the YNT_SSNG simulation. We also find that the correlations are slightly lower and RMS errors are slightly higher 212 in the YNT_SSNG simulation than in the <u>AP_XM_obsAP-XMEPA</u> simulation. 213

214 Observed HCHO shows peak amounts in excess of 4 ppbv (black lines, Figure 4) which are underestimated in both 215 simulations (red lines, Figure 4). However, the YNT_SSNG simulation tends to have overall higher HCHO mixing 216 217 218 220 221 222 223 224 225 226 227 228 ratios then the AP_XM_obsAP-XMEPA simulation leading to a nearly 50% reduction (-0.26 versus -0.543.4-ppbv) in the low bias relative to the EPA measurements. This is in spite of the fact that the YNT_SSNG uses a more realistic, and lower (relative to climatology) Green Vegetation Fraction (see Figure 2 in Otkin et al, 2023) which would tend to reduce biogenic VOC emissions. This suggests that anthropogenic VOC emissions may be playing a role in the reduction of the low biases in the YNT_SSNG simulation. This is likely due to the incorporation of more realistic (relative to climatology) GVF observations in the YNT_SSNG simulation and better representation of biogenic VOCs. Compared to the AP XM_obsAP-XMEPA simulation, we also find correlations and RMS errors are slightly lower in the YNT_SSNG simulation.

The larger high biases in NO2 and reduced low biases in HCHO in the YNT_SSNG simulation leads to significant reductions in high biases in ozone in the YNT_SSNG compared to the AP-XM simulation (0.07 versues 1.76 ppbv, not shown) and may be due to more nighttime ozone titration in the YNT_SSNG simulation.

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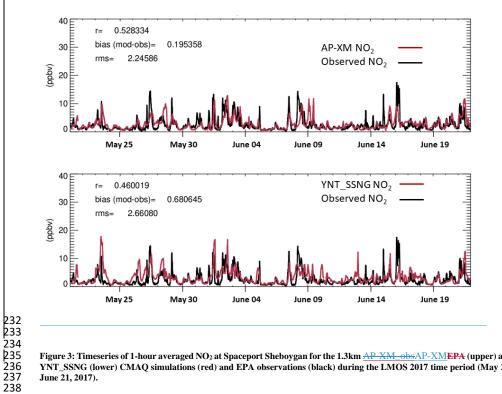
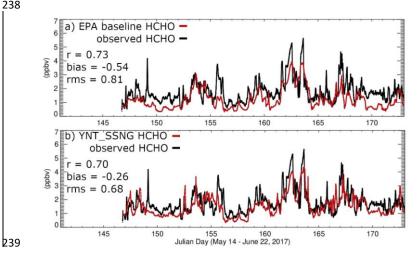
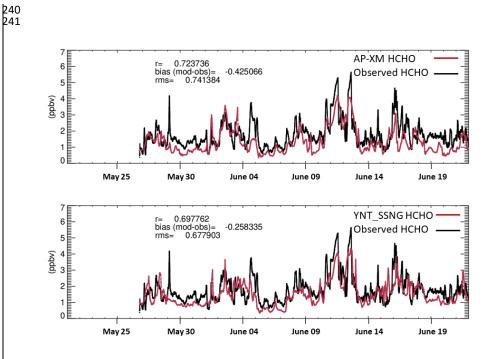


Figure 3: Timeseries of 1-hour averaged NO₂ at Spaceport Sheboygan for the 1.3km <u>AP_XM_obs</u>AP-XM<mark>EPA</mark> (upper) and YNT_SSNG (lower) CMAQ simulations (red) and EPA observations (black) during the LMOS 2017 time period (May 22-June 21, 2017).





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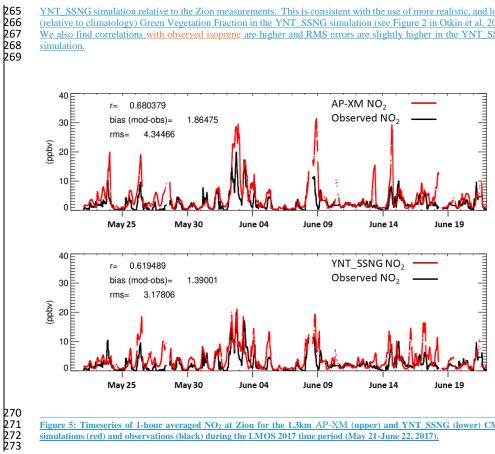
Figure 4: Timeseries of 1-hour averaged HCHO at Spaceport Sheboygan for the 1.3km <u>AP-XM-obs</u>AP-XMEPA (upper) and YNT_SSNG (lower) CMAQ simulations (red) and EPA observations (black) during the LMOS 2017 time period (May 245 22-June 21, 2017).

246 3.1.3 Evaluation with Zion IL ground-based NO₂ and iIsoprene measurements

247 248 249 250 251 252 253 254 255 256 257 258 259 260 261 262 263 264 During LMOS, the University of Wisconsin deployed a Thermo Scientific NO-NO2-NO2 Analyzer Model 42i instrument measuring in-situ NO2 and the University of Minnesota deployed a Proton-Transfer Quadrupole Interface Time-Of-Flight Mass Spectrometer (PTR-QiTOF) measuring ilsoprene at the LMOS Zion ground site to characterize ozone precursors along the shore of Lake Michigan. These 1-min measurements were co-located at the Illinois Air Monitoring site (17-097-1007) in Illinois Beach State Park, which is approximately 4 km south of the Chiwaukee monitor highlighted in Figure 2. Here, we use the hourly averaged NO_2 and isoprene measurements to evaluate the accuracy of prediction of ozone precursors at Zion for the YNT_SSNG and AP-XM CMAQ simulations.

Figures 5 and 6 show the hourly NO2 and isoprene comparisons, respectively. There are several periods where observed NO₂ (black lines, Figure 5) is above 10 ppbv; these periods are generally overestimated by the AP-XM simulation with YNT_SSNG simulation in much better agreement with observations (red lines, Figure 5). We find an overall positive bias of 1.86 ppbv for the AP-XM and an overall positive bias of 1.39 pbbv for the YNT_SSNG simulation. We also find that the correlations are slightly lower and RMS errors are lower in the YNT SSNG simulation than in the AP-XM simulation.

Observed il-soprene shows peak amounts in excess of 4 ppbv (black lines, Figure 6) which are significantly underestimated in both simulations (red lines, Figure 6). The YNT_SSNG simulation tends to have overall lower Hsoprene mixing ratios then the AP-XM simulation leading to a larger low bias (-0.34 versus -0.28 ppbv) for the



YNT_SSNG simulation relative to the Zion measurements. This is consistent with the use of more realistic, and lower (relative to climatology) Green Vegetation Fraction in the YNT_SSNG simulation (see Figure 2 in Otkin et al, 2023). We also find correlations with observed isoprene are higher and RMS errors are slightly higher in the YNT_SSNG

Figure 5: Timeseries of 1-hour averaged NO₂ at Zion for the 1.3km AP-XM (upper) and YNT_SSNG (lower) CMAQ simulations (red) and observations (black) during the LMOS 2017 time period (May 21-June 22, 2017).

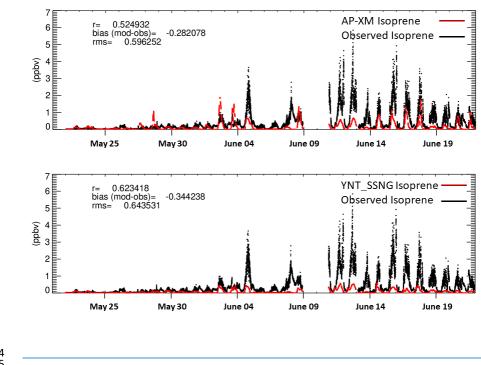


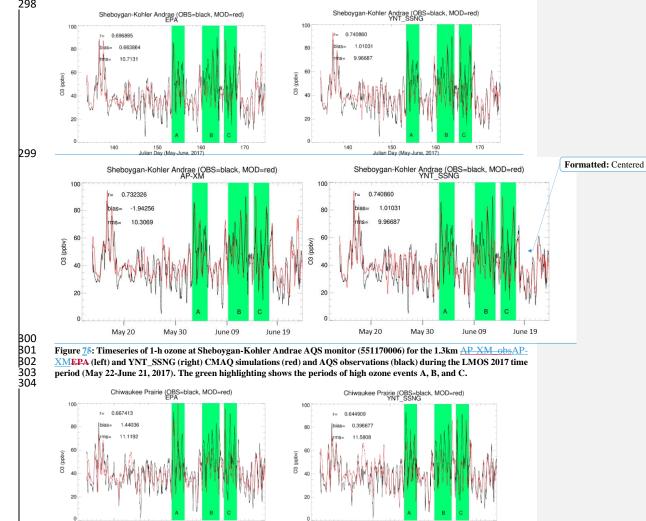
Figure 6: Timeseries of 1-hour averaged Hsoprene at Zion for the 1.3km AP-XM (upper) and YNT_SSNG (lower) CMAQ simulations (red) and observations (black) during the LMOS 2017 time period (May 21-June 22, 2017).

279 3.2 Model performance during high-ozone events

280 3.2.1 Sheboygan KA and Chiwaukee Prairie 1-hour ozone

In this and the following sections, we focus on the two AQS monitors that showed high ozone events during LMOS most clearly: the Sheboygan Kolher Andrae (KA) monitor (AQS 551170006), located south of Sheboygan, WI, and the Chiwaukee Prairie monitor (AQS 550590019) is located near the Wisconsin/Illinois border. These two sites are indicated by red circles in Figure 2.

Figures <u>75</u> and <u>86</u> show the hourly AQS observed and CMAQ <u>AP_XM_obsAP-XMEPA</u> and YNT_SSNG simulated
G₃ for Sheboygan KA and Chiwaukee Prairie monitors. Comparisons with AQS observations and the two simulations at Sheboygan KA show <u>similar increased</u> correlations (0.7<u>44</u> versus 0.<u>73697</u>), <u>reducedincreased</u> biases (1.01 versus <u>1.90.664</u> ppbv) and <u>similar reduced</u>-RMSE (9.97 versus 10.<u>3</u>7 ppbv) for the YNT_SSNG <u>relative to the AP-XM</u>
simulation. Similar comparisons at Chiwaukee Prairie show decreased correlations (0.64 versues 0.<u>7067</u>), <u>higherreduced</u> biases (0.<u>4397</u> versues <u>-0.13</u>+.44 ppbv) and increased RMSE (11.58 versues <u>11.5842</u> ppbv) for the YNT_SSNG relative to the AP-XM simulation. Student T-Tests between the AP-XM and YNT_SSNG simulations at each site show that the simulations have statistically significant differences (99% confidence level) in mean ozone concentration at Sheboygan KA but not at Chiwaukee Prairie. While the overall hourly ozone statistics at Sheboygan KA and Chiwaukee Prairie relatively similar between the <u>AP-XM_obsAP-XMEPA</u> and YNT_SSNG simulations



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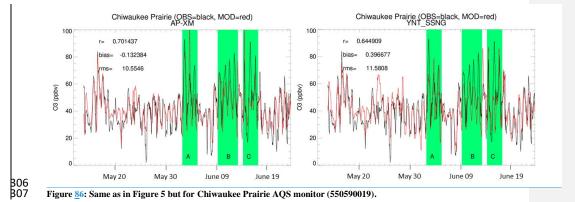
at these sites, the simulations during high ozone events are quite different. This is illustrated by looking at composite statistics during events A, B, and C.

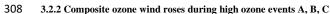
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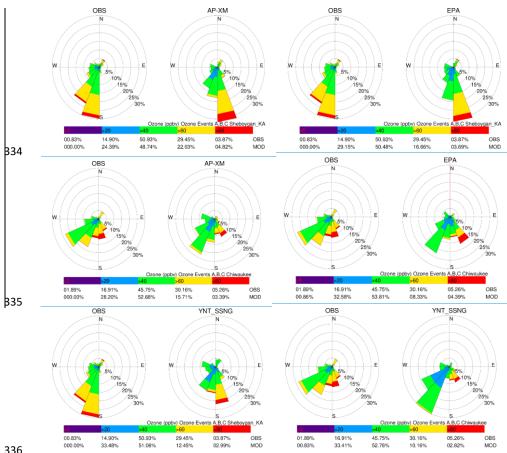
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809 Figure 97 shows observed and simulated composite ozone wind roses from the 1.3km AP-XM-obsAP-XMEPA and 310 YNT_SSNG simulations at the Sheboygan KA and Chiwaukee Prairie monitors during high ozone events A, B, and B11 B12 C. At Sheboygan KA, the observed wind direction is most frequently (>50%) from the south-southwest (SSW), which is also the direction where the majority of the higher (>60ppbv) ozone is observed during south southeast (SSE) ozone 813 814 events. The AP-XM-obsAP-XMEPA simulation predicts winds which are most frequently (>30%) from the southsoutheast (SSE) with the majority of the higher ozone coming from this direction. The YNT_SSNG simulation predicts 315 winds which are more variable but also most frequently (>20%) from the SSE with most of the higher ozone coming B16 from this direction. The overall frequency of higher ozone in the AP-XM-obsAP-XMEPA simulation (~270%) is 317 closer to the observed percentage (~33%) than the YNT_SSNG simulation (~15%). These comparisons show that the 818 AP-XM-obsAP-XMEPA meteorology best captures the observed ozone wind rose at Sheboygan KA during high 319 ozone events. 320

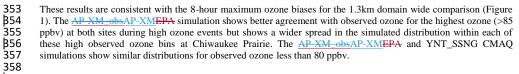
321 At Chiwaukee Prairie the observed winds are more variable and are most frequently (>40%) from the southwest. 322 While some of the observed higher ozone comes from the southwest, the highest (> 80 ppbv) ozone comes from the B23 B24 SSE. Both the <u>AP-XM_obsAP-XMEPA</u> and YNT_SSNG simulations frequently predict southwest winds (~30% and ~50%, respectively) with lower ozone (< 60 ppbv) than observed. Both the <u>AP-XM_obsAP-XMEPA</u> and YNT_SSNG B25 simulations show the highest ozone coming from the SSE, but the AP-XM YNT_SSNG simulation more accurately B26 B27 captures the observed percentages of high ozone coming from the SSE while AP_XM_obsEPA simulation shows both and high ozone coming from the southeast at Chiwaukee Prairie. The overall frequency of higher ozone in the 328 AP-XM-obsAP-XMEPA (~19%) and YNT_SSNG (~13%) are both lower than and YNT_SSNG simulations are very 829 similar (~13%) and both underestimate the observed percentage (~35%), however, most of the higher ozone in the 830 AP-XM-obsEPA simulation comes from the southeast. These comparisons show that the AP-XM_YNT_SSNG 331 simulation best captures the observed ozone wind rose at Chiwaukee Prairie during high ozone events but that both 332 simulations have a low bias in ozone when winds are from the southwest.

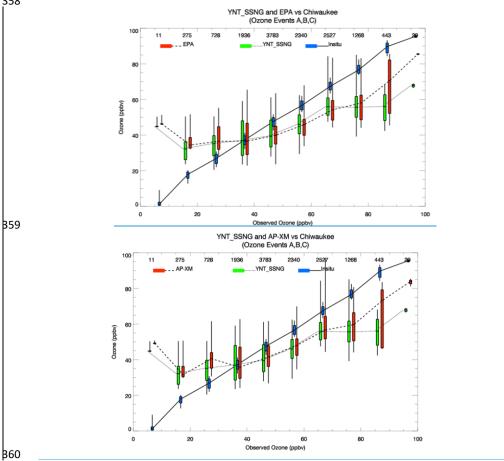


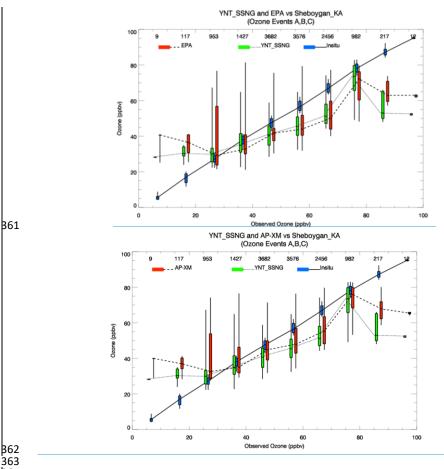
336 337 B38 339 Figure 97: Observed (OBS) and simulated wind roses using 1-h ozone and wind directions at the Sheboygan-Kohler Andrae AQS monitor (551170006, left columns) and Chiwaukee Prairie AQS monitor (550590019, right columns) for the 1.3km β40 341 AP XM_obsAP-XMEPA (upper rows) and YNT_SSNG (lower rows) CMAQ simulations during high ozone events A, B, and C. Wind directions are divided into 22.5 degree bins and the percentage of winds within each directional bin are indicted 342 343 344 by the percentages on the wind rose plots. The colors within each wind direction bin indicate the distribution of observed and simulated ozone within 20ppbv bins as indicated by the color bars. The overall percentage of observed (OBS) and simulated (MOD) ozone within each ozone bin is indicated below the color bar for each site and simulation.

345 3.2.3 1-h ozone concentration and wind direction during high ozone events A, B, C

846 847 348 849 850 While the ozone wind roses presented above provide a noverall comparison of the joint distribution of simulated and observed winds and ozone at these two stations, they do not provide a quantitative estimate of the errors in the simulations. In this section we have binned simulated and observed ozone and wind direction to provide a more quantitative characterization of the simulated biases. Figure 108 shows bar and whisker plots of 1.3km YNT_SSNG and AP-XM-obsAP-XMEPA CMAQ ozone simulations and 1-h averaged observed ozone at Chiwaukee Prairie and 351 Sheboygan KA during high ozone events A, B, and C. Both simulations show systematic high biases for lower 352 observed ozone concentrations (< ~40 ppbv) and low biases for higher ozone concentrations (> 50 ppbv) at both sites.







 B62
 Observed Ozone (ppb/)

 363
 Figure 108. Bar and whisker plots showing the binned median ozone concentrations from the 1.3km <u>AP-XM-obsAP-</u>

 864
 Figure 108. Bar and whisker plots showing the binned median ozone concentrations from the 1.3km <u>AP-XM-obsAP-</u>

 865
 XMEPA (dashed) and YNT_SSNG (dotted) CMAQ simulations and observed ozone (solid) at Chiwaukee Prairie (upper)

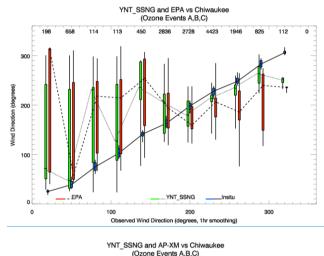
 366
 show the 95% for the <u>AP-XM-obsAP-XMEPA</u> (red) and YNT_SSNG (green) CMAQ simulations and observed ozone

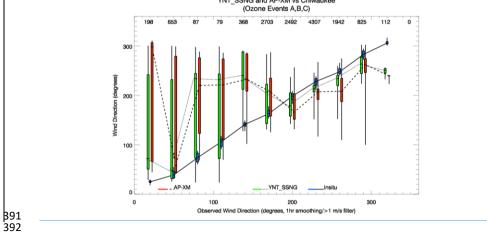
 867
 show the 95% for the <u>AP-XM-obsAP-XMEPA</u> (red) and YNT_SSNG (green) CMAQ simulations and observed ozone

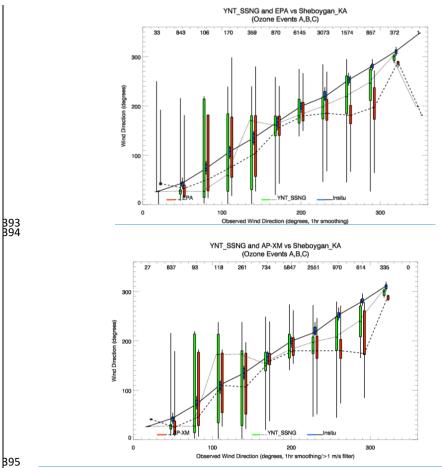
 868
 (blue). The total observed count within each 5 pbv bin is indicated on the top of each panel.

Figure 119 shows bar and whisker plots of 1.3km YNT_SSNG and <u>AP_XM_obs</u>AP-XMEPA CMAQ wind direction simulations and 1-h averaged observed wind direction for wind-speeds greater than 1 m/s at Chiwaukee Prairie and Sheboygan KA during high ozone events A, B, and C. <u>The 1 m/s threshold was included to reduce the impact of light</u>
and variable winds at these sites. Both simulations show a large westerly median bias and large variations and large variations in wind direction when the observed winds have an easterly component (0-180°) at Chiwaukee Prairie. This
could be associated with errors in the timing of the arrival of the lake breeze, but a more detailed analysis along the
lines of Wagner et al, (2022) would have to be performed to confirm thins. Winds with an easterly component account
for 30% of the observed wind directions at this site. Both simulations show a <u>smaller systematic</u> easterly bias in
median wind direction when the observed winds have a westerly component (180-360°) at-Chiwaukee Prairieboth
sites, but the YNT_SSNG simulation is in better agreement with observations <u>during these periods</u>. <u>when the wind is</u>
from the southwest, west, and northwest(240-290°). The two simulations show small (~20°) easterly biases when the

observed winds are from the prevailing wind direction at Chiwaukee (south to west southwest, 180 260°) and
 Sheboygan (south to south southwest, 180 220°) which account for 61% and 54% of the observed wind directions,
 respectively. The YNT_SSNG simulation shows the smallest (-10°) easterly biases during prevailing winds at
 Chiwaukee_The AP-XM simulation shows a small easterly biases when the observed winds have an easterly
 component at Sheboygan KA while the YNT_SSNG simulation still shows some westerly biases in median wind
 direction for these cases. Both simulations show somewhat larger easterly median biases when the observed winds better agreement with observations
 for these cases.







 B95
 Observed Wind Direction (degrees, 1hr smoothing/>1 m/s filler)

 B96
 Figure 119. Bar and whisker plots showing the binned median wind direction from the 1.3km <u>AP-XM-obsAP-XMEPA</u>

 (dashed) and YNT_SSNG (dotted) CMAQ simulations and observed wind direction (solid) at Chiwaukee Prairie (upper)

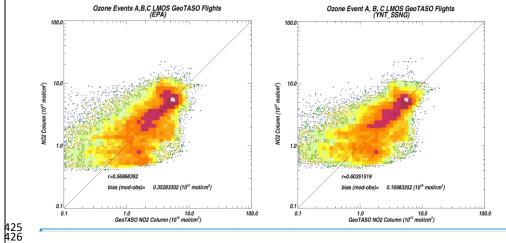
 and Sheboygan KA (lower) during high ozone events A, B, and C. The vertical bars show the 50% and the vertical lines

 show the 95% for the <u>AP-XM-obsAP-XMEPA</u> (red) and YNT_SSNG (green) CMAQ simulations and 1-hour averaged

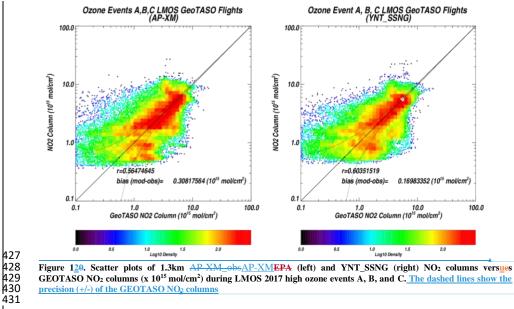
 observed wind direction (blue). The total observed count within each 20° bin is indicated on the top of the figures.

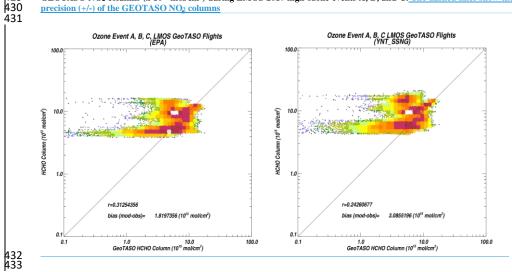
401 3.2.4 GEOTASO comparisons during high ozone events A, B, C

402 Here, we use GeoTASO (Nowlan et al., 2016) NO₂ and HCHO column measurements to verify ozone precursors 403 within the YNT_SSNG and <u>AP-XM_obsAP-XMEPA</u> simulations during high ozone events A, B, and C. Figure 1<u>2</u>0 404 shows the results of the NO₂ column analysis. Compared to observed NO₂ column measurements, the YNT_SSNG 405 and <u>AP-XM_obsAP-XMEPA</u> simulations have similar correlation (0.60 vs. 0.57) and the YNT_SSNG has a 406 substantially reduced bias (0.17 x 10¹⁵ vs. 0.3<u>1</u>5 x 10¹⁵ mol/cm²), <u>although still within the uncertainty of the GeoTASO 407 measurements, the YNT_SSNG has a lower correlation than the <u>AP-XM_obsAP-XMEPA</u> simulation (0.24 vs. 0.3<u>3</u>4) 408 and a larger bias (3.<u>1</u> θ x 10¹⁵ vs. 2.<u>3</u>4.<u>8</u> x 10¹⁵ mol/cm²).</u> 411 Nowlan et al. (2018) used comparisons between the GEOstationary Coastal and Air Pollution Events (GEO-CAPE) 412 Airborne Simulator (GCAS, which is similar to the GeoTASO instrument) NO2 and HCHO retrievals and columns 413 414 estimated from airborne in-situ NO2 and HCHO profiles to estimate mean precisions of 1 x 1015 mol/cm² and 19 x 10^{15} mol/cm² for the native (250m) resolution NO₂ and HCHO retrievals, respectively. The LMOS 2017 GeoTASO 415 radiances were co-added onto a 1km grid during the 2017 LMOS campaign so we anticipate that the precision of the 416 417 1km retrievals are better by a factor of 2. —Given the relatively high precision of GeoTASO NO₂ compared to the column amounts observed during high ozone events A, B and C, we conclude that the high bias in NO₂ columns in 418 the AP_XM_obsAP-XMEPA simulation is -significant, with more AP-XM NO2 columns found outside the estimated 419 420 $\pm -0.5 \times 10^{15}$ mol/cm² precision range then found in the YNT_SSNG simulation meaningful. We have less confidence in the significance of the differences between the YNT_SSNG and $\frac{AP_{\text{-}XM_{-}obs}AP_{\text{-}XM_{-}DS}AP_{\text$ 421 to the GeoTASO retrievals since the observed HCHO columns are on the order of the precision of the instrument (10 422 423 $x = 10^{15} \text{ mol/cm}^2$ and the biases in the column HCHO simulations measurements are both mostly less than the GeoTASO precision during high ozone events A, B, and C. <u>Overall, o</u>Our results show the YNT_SSNG simulation 424 has an improved representation NO2, which is a primary ozone precursor, during these high ozone events.



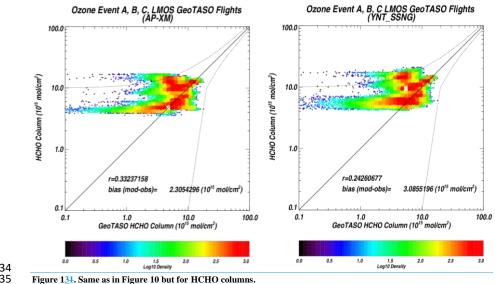
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437 3.3 June 2, 2017 ozone exceedance event

438 The only ozone exceedance event that had significant inland penetration of the lake breeze at both Chiwaukee and 439 Sheboygan KA during LMOS 2017 occurred on June 2nd, 2017 (Stanier et al, 2021; Wagner et al, 2022). - The 440 simulations on this day most clearly illustrates the differences between the AP_XM_obsAP-XMEPA and YNT_SSNG 441 442 results. Figure 142 shows the observed visible (0.64 micron) reflectance from the Advanced Baseline Imager (ABI) on the NOAA GOES-16 satellite and surface ozone concentrations from the YNT_SSNG and AP-XM_obsAP-443 XMEPA simulations, respectively, at 22 GMTUTC (5pm CDT) on June 2, 2017. To delineate the simulated 444 continental convective and stable maritime boundary layers we also show where the YNT_SSNG and AP-XM_obsAP-445 XMEPA simulated PBL heights are >1 km (Figure 142, Red or Green lines in the ABI panel). These contours roughly 446 correspond to the western most edge of the simulated marine boundary layer and indicate the extent of the penetration 447 of the lake breeze circulation. The ABI visible reflectances clearly show where the stable marine boundary layer 448 suppresses the formation of fair-weather cumulus clouds, which form within the turbulent continental boundary layer 449 and are evident to the west of the YNT_SSNG 1-km PBL height contour. The YNT_SSNG simulation shows a more 450 extensive penetration of high ozone concentrations inland, in agreement with the extent of the penetration of the 451 452 marine boundary layer. In contrast, the AP_XM_obsAP_XMEPA simulation shows very little penetration of the stable marine boundary layer. This lake breeze penetration has a significant impact on the simulated surface ozone 453 distributions with deeper penetration of high ozone inland. While the YNT_SSNG simulation shows deeper 454 penetration of the lake breeze circulation, it also leads to somewhat lower surface ozone concentrations near the 455 shoreline leading to underestimates in the observed ozone concentrations at this time.

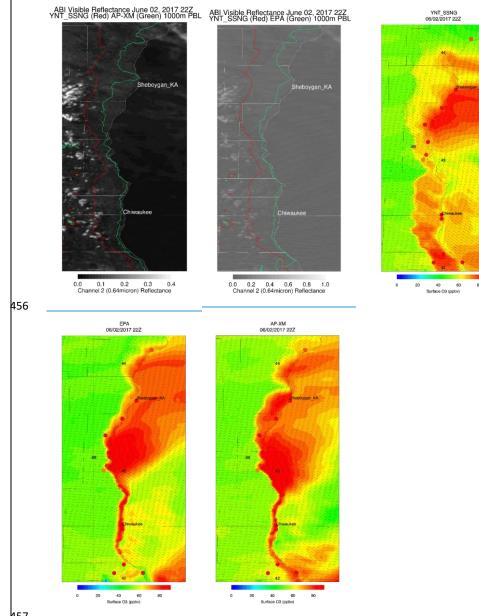




Figure 142: ABI visible (0.64micron) reflectance (left), YNT_SSNG surface ozone (ppbv, middle), and <u>AP_XM_obsAP_XM_bbsAP_</u> MEPA surface ozone (ppbv, right) at 22 <u>GMTUTC</u> on June 2, 2017. Observed AQS ozone concentrations at 22 <u>GMTUTC</u>

are shown as colored circles. Location of 1km YNT_SSNG (Red) and <u>AP_XM_obsAP-XMEPA</u> (Green) PBL heights are also shown in the ABI (left) panel. The locations of the Shebogan, KA and Chiwaukee Prairie AQS monitors are labeled in each of the panels.

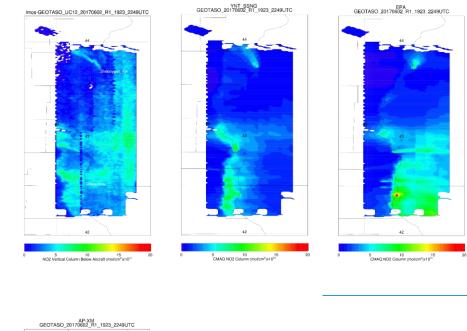
464 Figure 153 shows comparisons between airborne GeoTASO, YNT_SSNG, and AP_XM_obsAP-XMEPA column NO2 465 along the western shore of Lake Michigan on June 2, 2017. Comparisons with column NO₂ distributions provide a 466 means of comparing the fidelity of the lake breeze transport of ozone precursors during this high ozone event. 467 $Observed \ NO_2 \ columns \ peak \ near \ 10 \ x \ 10^{15} \ mol/cm^2 \ and \ shows \ penetration \ of \ the \ high \ NO_2 \ column \ amounts \ inland$ 468 by the lake breeze circulation, consistent with the ABI visible reflectances. The observed NO₂ columns also show 469 enhancements over the lake on the eastern part of the GeoTASO raster pattern that are best captured by the AP-XM 470 simulation. The GeoTASO NO₂ columns show peak amounts of 10 x 10¹⁵ mol/cm² and significant inland penetration 471 of higher NO2 columns over the southern portion of the flight track. The YNT_SSNG NO2 column shows similar peak 472 amounts and shows similar, but not as far inland, penetration of the high NO₂ columns. The AP XM-obsAP-XMEPA 473 NO₂ column shows localized NO₂ columns over 15×10^{15} mol/cm² along the Lake Michigan shoreline and does not 474 predict as much onshore penetration. The narrow plume of higher GeoTASO NO2 column extending to the northwest 475 from the coast north of the Sheboygan KA AQS monitor is a signature of the Edgewater coal-fired power plant. The 476 477 YNT_SSNG and AP-XM_obsAP-XMEPA simulations also show this plume, but the YNT_SSNG simulation does a better job of capturing the northwestward transport of the plume while the AP_XM_obsAP-XMEPA simulation shows 478 transport of this narrow plume to the north-northeast. 479

480 Figure 164 shows comparisons between airborne GeoTASO, YNT_SSNG, and AP_XM_obsAP-XMEPA column 481 HCHO along the western shore of Lake Michigan on June 2, 2017. HCHO columns vary much less spatially than NO2 482 483 columns, despite both anthropogenic and differing meteorologically driven biogenic VOC emissions influencing the former, given that the former is formed through VOC oxidation while the latter is primarily associated with 484 anthropogenic emissions. Both simulations capture the observed north-to-south positive gradient providing some 485 confidence in the larger scale gradients.- However, the GeoTASO HCHO measurements show values in excess of 10 486 x 10¹⁵ mol/cm² over Lake Michigan that are not captured in either the YNT_SSNG or AP-XM_obsAP-XMEPA 487 HCHO column simulations. Given the lower precision GeoTASO HCHO columns, the differences between the 488 YNT_SSNG and AP-XM_obsAP-XMEPA HCHO columns are difficult to quantify with these measurements. We 489 note large differences between simulated and observed GeoTASO NO₂ and HCHO over the eastern portion of 490 observations. These observations were collected later during the flight and therefore subject to larger uncertainties 491 related to the impact of stratospheric NO2 and ozone absorption interferences-resulting in a drift in the baseline 492 measurements. 493

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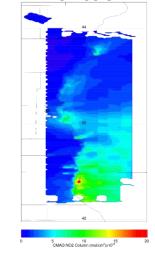
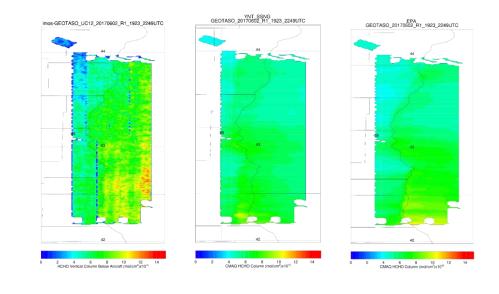


Figure 153: GEOTASO (left), YNT_SSNG (middle), and <u>AP-XM_obsAP-XM</u>EPA (right) Column NO₂ (x 10¹⁵ mol/cm²) on June 2, 2017. The location of the Sheboygan, KA AQS station is labeled in the GEOTASO column NO₂ panel.





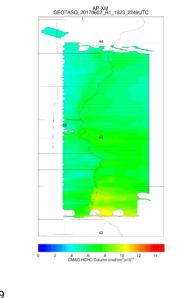


Figure 1<u>6</u>4: GEOTASO (left), YNT_SSNG (middle), and <u>AP-XM_obsAP-XM</u>EPA (right) Column HCHO (x 10¹⁵mol/cm²) on June 2, 2017.

503 Figure 175 shows comparisons between observed time height cross-sections of thermodynamic (temperature) and 504 kinematic (wind) distributions at Sheboygan WI during the June 02, 2017 ozone exceedance event. Observed 505 temperatures are obtained from the UW-Madison Atmospheric Emitted Radiance Interferometer (AERI) instrument 506 (Knutson et al, 2004a,b) while wind direction and speed are obtained from a Halo Photonics doppler wind lidar 507 instrument. Both of these instruments were deployed at the Sheboygan, WI ground site during LMOS 2017 (Stanier 508 et al, 2021; Wagner et al, 2022). AERI temperatures show a well-defined nocturnal boundary layer with a thin layer 509 of cold temperatures below 100-m AGL and a warmer layer extending up to approximately 600 m. The continental 510 convective boundary layer begins to form as the sun rises (~12 GMTZ [7am CDT]). This is evident in the warmer 511 512 513 surface temperatures near 15 GMTUTC (10am CDT). The AERI measurements show a new shallow layer of cooler air below 50m arriving at 17 GMTUTE (12pm CDT) associated with the stable marine boundary layer. Observed wind directions are out of the NW prior to 15 GMTUTC at 7m/s, rapidly diminish around 15 GMTUTC, and switch 514 515 to the SE around 18 GMTUTC (1pm CDT) when the lake breeze reaches Sheboygan, WI. Both simulations show an easterly bias during the observed NW winds, which is consistent with the overall statistics during ozone events A, B, 516 and C shown in Figure 11. The YNT_SSNG simulation captures the thermal structure of the nocturnal boundary layer 517 518 (temperature differences are less than 2°C below 100 m) and timing of the arrival of the maritime boundary layer but underestimates the near surface (below 200 m) convective boundary layer surface temperatures by up to 10°C within 519 the convective boundary layerat 15 GMT. The AP-XM simulation shows significant (temperature differences are 520 521 greater than 5°C below 100 m) overestimates of the nocturnal boundary layer temperatures and shows a gradual warming of temperatures below 200 m after 15 GMT, resulting in large (greater than 7,°C) overestimates in 522 523 524 temperatures and no evidence of the cooler lake breeze. Both simulations underestimate the observed increase in wind speed prior to the arrival of the lake breeze by ~2 m/s. The YNT_SSNG simulation shows a more captures rapid shift in wind direction associated with the arrival of the lake breeze than the AP-XM simulationthe vertical structure of the 525 526 527 lake breeze wind speed and direction, but the timing of the switch in wind direction is about 3 hours too early in the YNT_SSNG simulation. This results in errors in wind speeds of up to 5 m/s near 200 m in the YNT_SSNG simulation prior to the observed reduction in wind speed at 15 GMT. The observed depth of the wind shift is underestimated in 528 both simulations, but the YNT_SSN simulation does a better job of capturing the vertical extent of the wind shift and 529 reduction in wind speed above 200 m. This is most evident above 400 m where the AP-XM wind speeds are 530 underestimated by up to 5 m/s. In contrast, the AP XM_obsEPA simulation shows no thermodynamic signature of a 531 nocturnal or marine boundary layer and underestimates the sharp change in the observed windspeed and direction.

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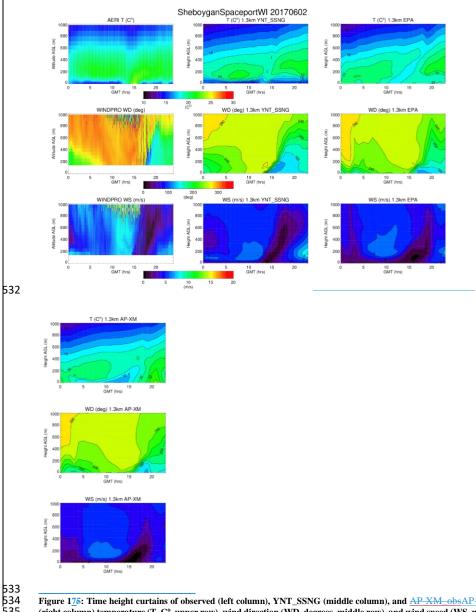




Figure 175: Time height curtains of observed (left column), YNT_SSNG (middle column), and <u>AP_XM_obsAP_XM</u>EPA (right column) temperature (T, C°, upper row), wind direction (WD, degrees, middle row), and wind speed (WS, m/s, lower row) at Sheboygan WI on June 2, 2017. Observed temperature is from UW-Madison AERI and observed winds are from the Halo Photonics doppler wind lidar instrument. 539 540 Figure 18 shows comparisons between observed time-height cross-sections of thermodynamic (temperature) and kinematic (wind) distributions at Zion IL during the June 02, 2017 ozone exceedance event. Observed temperatures 541 are obtained from the Microwave Radiometer while wind direction and speed were observed using a Sound Detection 542 543 and Ranging (SODAR) instrument, both of which were provided by the University of Northern Iowa. Both of these instruments were deployed at the Zion, IL ground site during LMOS 2017 (Stanier et al, 2021; Wagner et al, 2022). 544 Microwave temperatures show a well-defined nocturnal boundary layer with a thin layer of cooler temperatures below 545 100 m, similar to Sheboygan, WI (Figure 17) but not as cold. The continental convective boundary layer begins to 546 form as the sun rises (~12 GMT; -{7am CDT}). This is evident in the warmer surface temperatures near 15 GMT 547 (10am CDT). In contrast to Sheboygan, the Microwave temperatures do not show a signature of the cooler air 548 associated with the stable marine boundary layer. This may be due to the fact that the Zion site is further inland than 549 the Sheboygan site and turbulent heat fluxes from the warmer land surface warm the marine layer. The SodarODAR 550 wind direction shows a sharp transition from south-westerly to south-easterly winds and a rapid reduction in wind 551 speed (from over 10 m/s to less than 5 m/s) at 15 GMT associated with the arrival of the lake breeze at the Zion site. 552 Both the YNT_SSNG and AP-XM simulations overestimate the temperature within the nocturnal boundary layer with 553 the AP-XM showing somewhat larger (>5°C) warm biases (>5°C) below 100 m compared to the YNT_SSNG (<5°C) 554 below 100m. The YNT_SSNG simulation captures the development of the continental convective boundary layer 555 better than the AP-XM simulation, which underestimates the observed temperatures by 5-7°C below 100 m between 556 sunrise (12 GMT) and 15 GMT (10am CDT). This cold bias persists until 20 GMT (3pm CDT) in the AP-XM 557 simulation. The YNT_SSNG simulation shows some evidence of a cooler lake breeze moving over the Zion site that 558 leads to a cold bias of 5-7°C at 20 GMT. The YNT_SSNG simulation does a very good job in capturingmore accurately 559 captures the timing of the wind shift at 15 GMT, which is delayed by nearly 3 hours in the AP-XM simulation. Wind 560 speeds are similar in both simulations, although the YNT_SSNG simulation shows slightly higher (8 m/s versues 7 561 m/s) wind speeds prior to the arrival of the lake breeze and stronger (6 m/s versues 3 m/s) low level winds after 20 562 GMT, which are in better agreement with the SODARodar measurements.

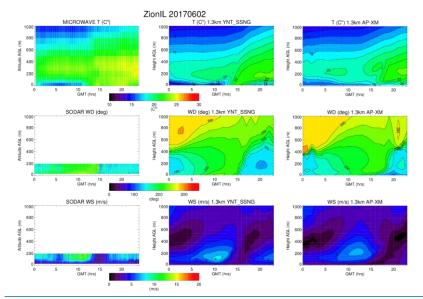


Figure 18: Time height curtains of observed (left column), YNT_SSNG (middle column), and AP-XM (right column) temperature (T, Co, upper row), wind direction (WD, degrees, middle row), and wind speed (WS, m/s, lower row) at Zion Ill on June 2, 2017. Observed temperatures were obtained using a Microwave Radiometer and observed winds are from a SODAR instrument provided by Alan Czarnetski at the University of Northern Iowa.

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569 4. Discussion and Conclusions

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570 We have conducted an evaluation of two model simulations employing differing meteorological inputs, with the goal 571 of identifying a model configuration best suited for characterizing the spatial and temporal variability of ozone and its 572 precursors where lake breezes commonly affect local air quality along the Lake Michigan shoreline. We focus on the 573 period of the LMOS campaign, 22 May - 22 June 2017, using the innermost grid of a triple-nested simulation around 574 575 Lake Michigan, with a horizontal resolution of 1.3 km. The AP-XM_obsAP-XMEPA simulation used the same boundary layer and surface physics that are used within CMAQ_best-practice recommendations (Personal 576 communication, Jon Pleim and Robert Gilliam US EPA) for WRF inputs to the CMAQ model, including nudging to 577 578 ved near surface conditions; our YNT SSNG simulation used different WRF parameterizations, as well as constraints to satellite observations of green vegetation fraction and soil temperature and moisture, as detailed by 579 Otkin et al. (20232). 580

581 Both model simulations reasonably capture observed daily maximum 8-h average ozone amounts over the study 582 period, however both simulations underestimateding ozone amounts at times with high ozone and overestimateding 583 ozone when observed amountsations were lower. These ozone biases are consistent with those simulated by Baker et 584 al. (2023). Both model simulations also perform similarly on an hourly basis on high ozone days. We find the AP-585 XM_obsAP-XMEPA simulation better represents hourly ozone when observed amounts are high (80-90 ppbv), and 586 the YNT_SSNG simulation overall biases are generally smaller (less negative) than those of the AP-XM_obsAP-587 XMEPA simulation. Both simulations also tend to underestimate amounts of the ozone precursor HCHO, with smaller 588 (less negative) biases in the YNT_SSNG simulation. This is inspite of the factdespite that the YNT_SSNG uses a 589 more realistic, and lower (relative to climatology) Green Vegetation Fraction (which would tend to reduce biogenic 590 VOC emissions) suggesting that anthropogenic HCHO emissions may be playing a more important role in HCHO 591 concentrations in the YNT_SSNG simulation. Both simulations significantly underestimate isoprene, with larger 592 (more negative) biases in the YNT SSNG simulation. This is also consistent with the use of more realistic, and lower 593 (relative to climatology) Green Vegetation Fraction in the YNT_SSNG simulation. -likely-resulting from the 594 595 incorporation of satellite derived GVF and therefore more realistic in line calculations of biogenic VOC emissions. We find the simulations are less similar in their representation of NO₂ amounts; while the AP-XM_obsAP-XMEPA 596 simulation tends to underestimate NO2 at monitor sites, the YNT SSNG simulation has an overall high (positive) bias. 597

598 Since the simulations use identical anthropogenic_emissions and chemistry, the differences in modeled ozone and 599 precursors are_linked to anythe differences in biogenic emissions resulting from the input meteorology and from 600 differences in boundary layer mixing, both-horizontal and vertical transport. In modeling the same time period, Baker 601 et al. (2023) In Part 1 of this study, Otkin et al. (2022) noted the AP-XM-obsAP-XMEPA simulation had an overall 602 low bias in wind speed, the YNT_SSNG simulation had a positive bias, and the simulations had similar RMSE. Here, 603 we find many similarities between simulations on high ozone days. At Chiwaukee Prairie, we find both simulations 604 capture the highest ozone amounts transported from the SSE. On high ozone days at Sheboygan KA, observed winds 605 tend to be SSW, while both models show highest ozone amounts transported from the SSE. At Chiwaukee Prairieboth 606 locations, both simulations tend to have a westerly bias when observed winds have an easterly (onshore) component. 607

608 We find greater differences in column amounts of ozone precursors. The AP-XM-obsAP-XMEPA simulation has a 609 negative bias in near-surface NO2 at Sheboygan, a high bias in near-surface NO2 at Zion, -a positive bias in column 610 NO₂ amounts, and elevated column amounts concentrated along the Lake Michigan shoreline during the ozone 611 exceedance event on June 2nd. The YNT_SSNG simulation has a small positive bias in NO₂ column amounts, with 612 elevated column amounts extending further inland on the lake-breeze enhanced ozone event on June 2nd. While these 613 differences reflect the parameterizations used to generate input meteorology, differences in vertical mixing, and 614 ensuing column amounts of NO2 and HCHO discussed here, they are further complicated by CMAQ using the ACM2 615 parameterization for vertical diffusion-a mismatch for the YNT_SSNG simulation that influences our evaluation 616 since it leads to differences in boundary layer mixing. Still, the NO₂ column comparisons provide support for the 617 improved representation of lake breeze transport of ozone precursors in the YNT_SSNG simulation. Future model 618 comparisons with upcoming geostationary observations will allow for maturing analysis for assessing model 619 performance with respect to the diurnal evolution of precursors during ozone events.

Our thermodynamic and kinematic comparison of the <u>AP-XM-obsAP-XMEPA</u> and YNT_SSNG simulations show
 improved representation of not only the extent, but of the timing of the lake breeze in the YNT_SSNG simulation at
 <u>Sheboygan, WI and Zion, IL for the June 2, 2017 ozone episode</u>. This is consistent with the meteorological analysis

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presented in Part 1 of this study, where the YNT_SSNG had a better representation of diurnal patterns (Otkin et al., 2023²). However, we note that the meteorological inputs to our CMAQ simulations are hourly, as is typically used in air quality modeling studies. Both simulations would likely benefit from sub-hourly winds given the rapid changes that can occur in the presence of lake and land breeze circulations. For this, <u>a more tightly a two-way</u>, coupled model such as WRF-CMAQ (Wang et al., 2021) would be better suited for a goal of better simulating the fine temporal and spatial scales of lake breeze transport and chemistry.

631 This analysis complements other studies in evaluating the impact of changing meteorological inputs and 632 parameterizations on air quality in a complex environment. Appel et al. (2014) also found improved representation of 633 ozone in environments with bay and sea breezes with the addition of high-resolution SST into the WRF and CMAQ 634 modeling framework. Cheng et al. (2012) underscored the importance of PBL parameterization in simulating land-sea 635 breezes and their impacts on near-surface ozone. Similar to our work, Banks and Baldasano (2016) evaluated the 636 impacts of PBL parameterizations on air quality and also found ambiguous results, with the simulation using the YSU 637 PBL better capturing observed NO2, and the simulation using the ACM2 PBL better capturing observed ozone. Future 638 work will be able to take advantage of ongoing improvements to both WRF and CMAQ, such as an update to the 639 calculation of vegetative fraction and PX-LSM soil parameters in WRF (Appel et al., 2021), and should explore the 640 relationships among spatial and temporal resolution of meteorological parameterizations themselves along with those 641 of the modeling framework. 642

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