



1 Disappearing day-of-week ozone patterns in US nonattainment areas

2 Heather Simon¹, Christian Hogrefe², Andrew Whitehill², Kristen M. Foley², Jennifer Liljegren³,
3 Norm Possiel¹, Benjamin Wells¹, Barron H. Henderson¹, Lukas C. Valin², Gail Tonnesen⁴, K.
4 Wyatt Appel², Shannon Koplitz¹

5
6 ¹US Environmental Protection Agency, Office of Air and Radiation, Research Triangle Park, NC

7 ²US Environmental Protection Agency, Office of Research and Development, Research Triangle Park, NC

8 ³US Environmental Protection Agency, Region 5, Chicago, IL

9 ⁴US Environmental Protection Agency, Region 8, Denver, CO

10 *Correspondence to:* Heather Simon (Simon.Heather@epa.gov)

11

12 **Abstract.** Past work has shown that traffic patterns in the US and resulting NO_x emissions vary by day of week, with
13 NO_x emissions typically higher on weekdays than weekends. This pattern of emissions leads to different levels of
14 ozone on weekends versus weekdays and can be leveraged to understand how local ozone formation changes in
15 response to NO_x emissions perturbations in different urban areas. Specifically, areas with lower NO_x but higher ozone
16 on the weekends (the weekend effect) can be characterized as NO_x-saturated and areas with both lower NO_x and
17 ozone on weekends (the weekday effect) can be characterized as NO_x-limited. In this analysis we assess ozone
18 weekend-weekday differences across US nonattainment areas using 18 years of observed and modeled data from
19 2002-2019 using two metrics: mean ozone and percentage of days > 70 ppb. In addition, we quantify the modeled and
20 observed trends in these weekend-weekday differences across this period of substantial NO_x emissions reductions in
21 the US. The model assessment is carried out using EPA's Air QUALity TimE Series Project (EQUATES) CMAQ
22 dataset. We identify 3 types of ozone trends occurring across the US: disappearing weekend effect, disappearing
23 weekday effect, and no trend. The disappearing weekend effect occurs in a subset of large urban areas that were NO_x
24 -saturated (i.e., VOC-limited) at the beginning of the analysis period but transitioned to mixed chemical regimes or
25 NO_x-limited conditions by the end of the analysis period. Nine areas have disappearing weekend effect trends in both
26 datasets and with both metrics indicating strong agreement that they are shifting to more NO_x-limited conditions:
27 Milwaukee, Houston, Phoenix, Denver, Northern Wasatch Front, Southern Wasatch Front, Las Vegas, Los Angeles –
28 San Bernardino County, Los Angeles – South Coast, and San Diego. The disappearing weekday effect was identified
29 for multiple rural and agricultural areas of California which were NO_x-limited for the entire analysis period but appear
30 to become less influenced by local day of week emission patterns in more recent years. Finally, we discuss a variety
31 of reasons why there are no statistically significant trends in certain areas including complex impacts of heterogeneous
32 source mixes and stochastic impacts of meteorology. Overall, this assessment finds that the EQUATES modeling
33 simulations indicate more NO_x-saturated conditions than the observations but do a good job of capturing year-to-year
34 changes in weekend-weekday ozone patterns.

35

36 **1 Introduction**

37 Ground-level ozone (O₃), a key component of photochemical smog, has adverse impacts on human health and
38 ecosystems (U.S. Environmental Protection Agency, 2019). In the United States (US), the Clean Air Act Amendments
39 of 1970 instruct the Environmental Protection Agency (EPA) to set National Ambient Air Quality Standards



40 (NAAQS) for criteria pollutants. Since 1979, O₃ has served as the indicator species for the criteria pollutant of
41 photochemical oxidants (44 FR 8202) and since 1997, the form of the standard has been determined by the 3-year
42 average of the annual 4th-highest daily maximum 8-hour concentration (MDA8) (62 FR 38856). In 2015, the O₃
43 NAAQS were revised to the current level of 0.070 ppm or 70 ppb (80 FR 65291). As of 2018, 52 areas in the US had
44 been designated as nonattainment of the 2015 O₃ NAAQS (83 FR 25776; 83 FR 35136; 83 FR 52157).

45

46 O₃ is predominantly a secondary pollutant formed from photochemical reactions of nitrogen oxides (NO_x) and volatile
47 organic compounds (VOCs). Ground-level O₃ concentrations are a complex nonlinear function of the chemistry of
48 natural and anthropogenic precursor emissions, as well as meteorology, transport, and deposition (Seinfeld and Pandis,
49 2016). O₃ formation rates depend on the concentrations and speciation of NO_x and VOCs. To reduce ambient O₃
50 concentrations, control strategies have been enacted in the US over the last 50 years to control the emissions of both
51 NO_x and VOCs (Simon et al., 2015).

52

53 The effectiveness of different control strategies on O₃ production rates depends on the photochemical environment
54 under which ozone is formed. Ozone formation environments are typically categorized as either NO_x-limited or NO_x-
55 saturated, with a mixed or transitional regime between the two (Sillman, 1995, 1999; Sillman et al., 1990). In the NO_x-
56 limited regime, ambient ozone concentrations will respond more strongly to changes in NO_x emissions than VOC
57 emissions. In contrast, in a NO_x-saturated (or VOC-limited) regime ozone will increase with NO_x emission controls
58 but will decrease with VOC emissions controls. Understanding the photochemical regimes of different ozone
59 nonattainment areas and how they have changed over time is important for understanding the impacts of previous
60 control strategies and guiding future control strategies to have the maximum health benefit with the least economic
61 burden.

62

63 Different methods have been proposed to determine ozone formation regimes and their changes over time. One
64 common method used to evaluate ozone formation chemistry is through day-of-week (DOW) differences in the
65 concentration of ozone and its precursors. The DOW effects leverage NO_x emissions differences between weekdays
66 and weekends (Marr and Harley, 2002a, b). In the US, onroad vehicles are a dominant source of NO_x emissions (Toro
67 et al., 2021). Diesel vehicle traffic tends to be higher on weekdays (Monday through Friday) than on weekends
68 (Saturday and Sunday). This results in higher NO_x emissions on weekdays than weekends (Marr and Harley, 2002a,
69 b). Daily varying emissions sources such as diesel vehicles are not a major source of VOC emissions. In addition,
70 VOC emissions in some areas are dominated by biogenic emissions that do not vary by day of week. Consequently,
71 VOC emissions are generally similar on weekends and weekdays in most areas. The result of DOW NO_x patterns is
72 that ozone concentrations tend to be higher on weekends than weekdays in NO_x-saturated areas and lower on
73 weekends than weekdays in NO_x-limited areas (Kopplitz et al., 2022). DOW differences in ozone were first reported
74 in the 1970s (Bruntz et al., 1974; Cleveland et al., 1974). In 2002 the DOW ozone differences in California were
75 explicitly tied to DOW patterns in diesel vehicle traffic (Marr and Harley, 2002a, b). Since that time, multiple studies
76 have used DOW ozone patterns to assess ozone chemical formation regimes in individual US cities including Los



77 Angeles, California (Chinkin et al., 2003; Fujita et al., 2003b; Fujita et al., 2003a; Gao, 2007; Gao and Niemeier,
78 2007; Warneke et al., 2013), Fresno, California (De Foy et al., 2020), Sacramento, California (Murphy et al., 2007),
79 Phoenix, Arizona (Atkinson-Palombo et al., 2006), Atlanta, Georgia (Blanchard and Tanenbaum, 2006), Baltimore,
80 Maryland (Roberts et al., 2022), and New York City, New York (Singh and Kavouras, 2022). A smaller number of
81 studies have assessed ozone DOW patterns across multiple US urban areas (Blanchard et al., 2008; Jaffe et al., 2022;
82 Koo et al., 2012; Koplitz et al., 2022; Pun et al., 2003). Additionally, ozone DOW patterns have been used as a method
83 for assessing chemical formation regimes outside of the US in Shanghai, China (Zhang et al., 2023), the Lesser Antilles
84 Archipelago (Plocoste et al., 2018), Rio de Janeiro, Brazil (Martins et al., 2015), Santiago, Chile (Rubio et al., 2011),
85 Andalusia, Spain (Adame et al., 2014), the Iberian Peninsula (Jiménez et al., 2005), Athens, Greece (Paschalidou and
86 Kassomenos, 2004) and in multiple other European cities (Pires, 2012). One complication with interpreting DOW O₃
87 patterns is that O₃ concentrations in urban areas are generally impacted by a mix of transport and local formation. O₃
88 transport can occur over a variety of timescales. In some locations there could be a regional O₃ DOW effect that might
89 be evident as a slightly lagged timescale depending on typical transport times from major upwind urban source areas.

90

91 Previous work has shown a substantial decrease in NO_x emissions in the US over the past 20 years as a result of
92 national, state, and local regulations (Krotkov et al., 2016; Lamsal et al., 2015; Russell et al., 2012; Toro et al., 2021).
93 Concurrent with the US NO_x decreases, multiple studies have found that ozone chemical formation regimes have also
94 changed in the US (Jin et al., 2020; Jin et al., 2017; Koplitz et al., 2022). In this paper, we focus on 51 areas in the US
95 which were designated in 2018 as nonattainment ([https://www.epa.gov/green-book/green-book-8-hour-ozone-2015-](https://www.epa.gov/green-book/green-book-8-hour-ozone-2015-area-information)
96 [area-information](https://www.epa.gov/green-book/green-book-8-hour-ozone-2015-area-information)) under the 2015 O₃ NAAQS (some of these areas have since been redesignated to attainment based
97 on clean monitoring data). We look at changes in DOW patterns in the US over 18 years from 2002 to 2019 using
98 both measured and modeled data to provide insights into how ozone formation chemistry has changed in the US as a
99 result of emissions reductions, and to assess how well modeling is able to capture the observed changes. This 18-year
100 dataset, which is part of EPA's Air QUALity Time Series Project (EQUATES), is unique in its application of consistent
101 emissions and modeling methodologies across the entire analysis period providing an opportunity to assess multi-year
102 trends.

103

104 **2 Methods**

105

106 For this assessment we use MDA8 ozone monitoring data obtained from EPA's Air Quality System (AQS)
107 (<https://www.epa.gov/aqs>) and MDA8 ozone modeling data from simulations of the Community Multiscale Air
108 Quality model version 5.3.2 (CMAQv5.3.2). The CMAQ model data are part of EQUATES which provides an 18-
109 year set of modeled meteorology, emissions, air quality and pollutant deposition spanning the years 2002 through
110 2019 using consistent modeling methods across years. The CMAQv5.3.2 model configuration, including input data,
111 boundary conditions, and science options are available from US EPA (Epa, 2021). The emissions inventories
112 developed for the EQUATES CMAQ modeling are described in (Foley et al., 2023).

113



114 We extract CMAQ modeling data only for days and grid-cells with monitoring data such that both datasets are paired
115 in time and location. Both datasets are subset to ozone monitors located within 51 of the 52 areas that were designated
116 in 2018 as nonattainment for the 2015 O₃ NAAQS (a list of areas is available in Tables S1 and S2) (83 FR 25776; 83
117 FR 35136; 83 FR 52157). Because this analysis focuses on May-September data, we do not include data from the
118 Uintah Basin nonattainment area for which violations of the NAAQS predominantly occur in winter months. Data are
119 analyzed for the 18-year period of the EQUATES modeling dataset.

120

121 We start by analyzing changes in ozone between weekends and weekdays pooled across all monitoring locations for
122 each nonattainment area for 5-year rolling periods (i.e., 14 different periods covering the 18-year timeseries). We pool
123 data into 5-year periods for several reasons. First, it dampens impacts of interannual meteorology that can contribute
124 to large year-to-year changes in ozone for a given location. Previous work has shown that differential meteorological
125 patterns on weekends versus weekdays impacts ozone DOW patterns in a single year and that pooling data across
126 multiple years can reduce this effect (Pierce et al., 2010). Second, it provides a larger sample size for calculating ozone
127 differences between weekends and weekdays. The use of 5-year periods does, however, limit this analysis' ability to
128 parse out changes in weekend-weekday differences that have occurred due to emissions changes in the most recent
129 individual years analyzed. For example, any changes occurring only in 2018 and/or 2019 would be dampened in the
130 2015-2019 pooled data.

131

132 For the purpose of quantifying differences in weekend versus weekday O₃ concentrations, we use Sundays to represent
133 weekends (WE) and Tuesdays, Wednesdays and Thursdays to represent weekdays (WD). We do not include ozone
134 on Monday and Saturday to minimize any carryover impacts on concentrations from the previous day and we exclude
135 Friday as it may exhibit somewhat different emissions patterns than the other weekdays.

136

137 We use two metrics to quantify differences in ozone between weekends and weekdays. First, we quantify mean
138 differences in ozone across the entire distribution of days in each season (Winter = Dec, Jan, Feb; Spring = Mar, Apr,
139 May, Summer = Jun, Jul, Aug, Fall = Sep, Oct, Nov, ozone season = May-Sep) using Eq. (1), where O_{3,WE} represents
140 MDA8 O₃ on Sundays and O_{3,WD} represents MDA8 O₃ on Tuesdays, Wednesdays, and Thursdays.

141

$$142 \quad \overline{\Delta O_{3,DOW}} = \overline{O_{3,WE}} - \overline{O_{3,WD}} \quad (1)$$

143

144 In this study we mainly focus on differences during the May-Sep ozone season. The Welch's t-test (Welch, 1947) is
145 used to denote whether the mean WE-WD difference is statistically different from zero ($p < 0.05$). All available ozone
146 monitoring data and model output from all monitoring locations within each nonattainment area are included in the
147 calculation, providing a measure of average behavior across each area. We also examine 24-hour average modeled
148 formaldehyde and NO_x concentrations at each of the ozone monitor locations to verify whether the model shows
149 expected patterns of higher NO_x on weekdays than on weekends and trends in these ozone precursors. Formaldehyde
150 is used as an indicator of first-generation VOC reaction products for this purpose. We note that monitoring data for



151 VOCs and NO_x are much sparser in terms of sampling frequency and spatial density than ozone measurements, so we
152 rely on the model alone to verify underlying day-of-week patterns in precursor compounds.

153

154 Second, similar to (Jaffe et al., 2022), we look at the percent of days with MDA8 ozone values above the NAAQS
155 level of 70 ppb. We calculate the percent of total weekends and weekdays in May-Sep for which MDA8 ozone
156 concentrations exceeded 70 ppb as shown in Eq. (2).

157

$$158 \Delta O_{3,DOW,\%>70} = O_{3,WE,\%>70} - O_{3,WD,\%>70} \quad (2)$$

159

160 For this calculation, a day is characterized as exceeding the NAAQS in an area if measured and/or modeled ozone is
161 above 70 ppb at the location of any ozone monitor within the area. In this way we are tracking days where some
162 portion of the area has observed or modeled ozone above 70 ppb, but the analysis does not distinguish whether the
163 high ozone concentrations are localized over a small portion of the area or widespread across multiple monitoring
164 locations. This analysis also does not consider whether days with modeled ozone above 70 ppb occur simultaneously
165 with observed ozone above 70 ppb. We use the Fisher's exact test (Fisher, 1935; Mehta and Patel, 1983) to determine
166 whether the proportion of days above 70 ppb differs significantly ($p < 0.05$) between weekends and weekdays.

167

168 Next, we use the Theil-Sen estimator (Sen, 1968; Theil, 1992) to determine the multi-year trends in $\overline{\Delta O_{3,DOW}}$ and
169 $\Delta O_{3,DOW,\%>70}$ for each area. This nonparametric approach was chosen due to the small sample size ($n=14$ 5-year
170 windows) and the fact that the Thiel-Sen estimator does not require any assumptions on the distribution of the
171 residuals. The Mann-Kendall test (Kendall, 1975; Mann, 1945) is used to determine the statistical significance of the
172 derived trends in WE-WD O₃ differences. For each derived trend, we also document the 95% confidence interval.

173 Finally, investigation of relationships between WE-WD O₃ and meteorological parameters used the meteorological
174 dataset developed by and described in (Wells et al., 2021).

175

176 **3 Results**

177

178 **3.1 NO_x and formaldehyde day-of-week patterns**

179

180 For all but one of the 51 areas, the model shows clear patterns of higher NO_x concentrations on weekdays than
181 weekends and relatively constant formaldehyde concentrations across May-Sep days for the entire 2002-2019 analysis
182 period. This is consistent with the underlying assumption in the ozone day-of-week analyses discussed above. Here
183 we describe examples of the NO_x and formaldehyde day of week patterns using the data for Denver, CO and Los
184 Angeles, CA to show typical patterns in large urban areas and Butte County, CA to show a typical pattern in a more
185 rural area in Figures 1, 2, and 3, respectively. The modeled WE-WD differences in NO_x concentrations are more
186 pronounced in large urban areas such as Los Angeles and Denver than in rural or agricultural areas such as Butte



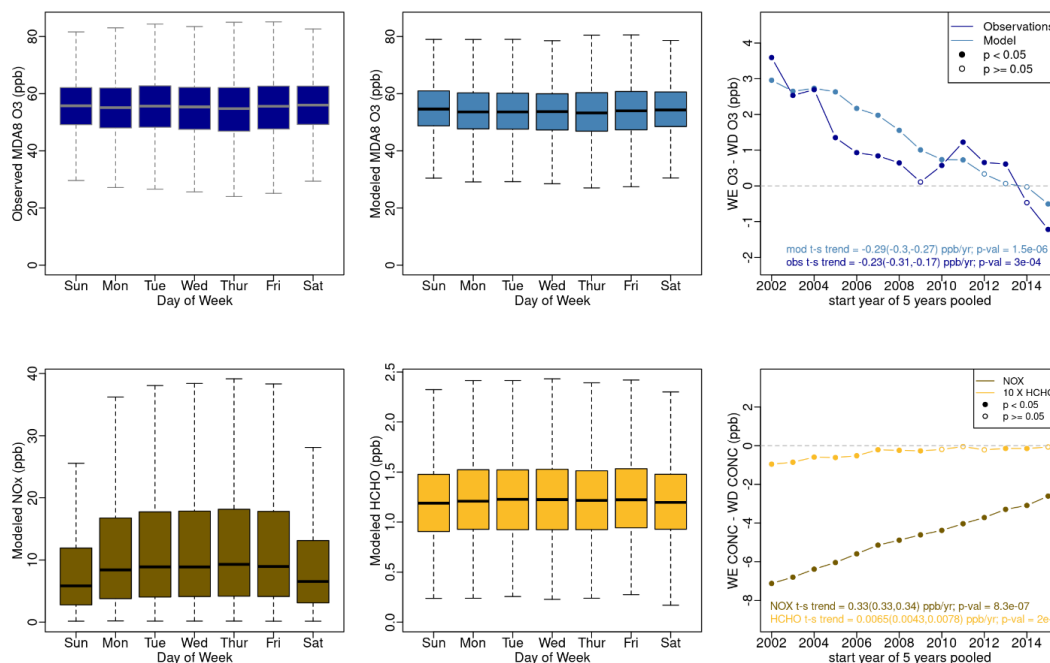
187 County. The only area that does not demonstrate higher modeled NO_x concentrations on weekdays than weekends is
 188 Door County, WI (Figure S-1). Higher NO_x emissions on weekdays are typically associated with commuting patterns
 189 and greater vehicular activity from commercial truck traffic. The nonattainment portion of Door County, which was
 190 fully redesignated to attainment in 2022 (87 FR 25410), is located at the tip of a peninsula on Lake Michigan and a
 191 rural recreation and tourist destination (i.e., likely to see more weekend activity). Consequently, the area does not
 192 follow typical weekday-weekend emission patterns and therefore modeled NO_x concentration patterns are unlike those
 193 of other areas. While the model does not predict substantial day-of-week formaldehyde differences in most areas,
 194 there are small modeled weekday formaldehyde enhancements in some areas such as Chicago (Figure S-2).

195

196 Theil-Sen trends show that differences in WE versus WD NO_x have diminished significantly over time in most areas
 197 (e.g. Figures 1, 2 and 3). The WE versus WD differences in formaldehyde are also diminishing over time but to a
 198 much lesser extent. As total emissions have decreased, absolute concentrations of NO_x have also decreased. Figures
 199 S-5 and S-6 show that the WE versus WD NO_x trends remain significant whether tracking absolute or normalized
 200 NO_x differences in Denver and Los Angeles, which is consistent with WE-WD NO_x trends seen in all but ten of the
 201 nonattainment areas. In nine areas (Houston, TX; Las Vegas, NV; Muskegon, MI; New York, NY; Phoenix, AZ; San
 202 Diego, CA; St. Louis, MO-IL; Tuolumne County, CA; and Yuma, AZ) while absolute WE-WD NO_x differences have
 203 diminished significantly there is no significant trend in relative WE-WD differences. In Mariposa County, CA neither
 204 absolute nor relative WE-WD NO_x differences have significant trends from 2002-2019.

205

Denver Metro/North Front Range, CO

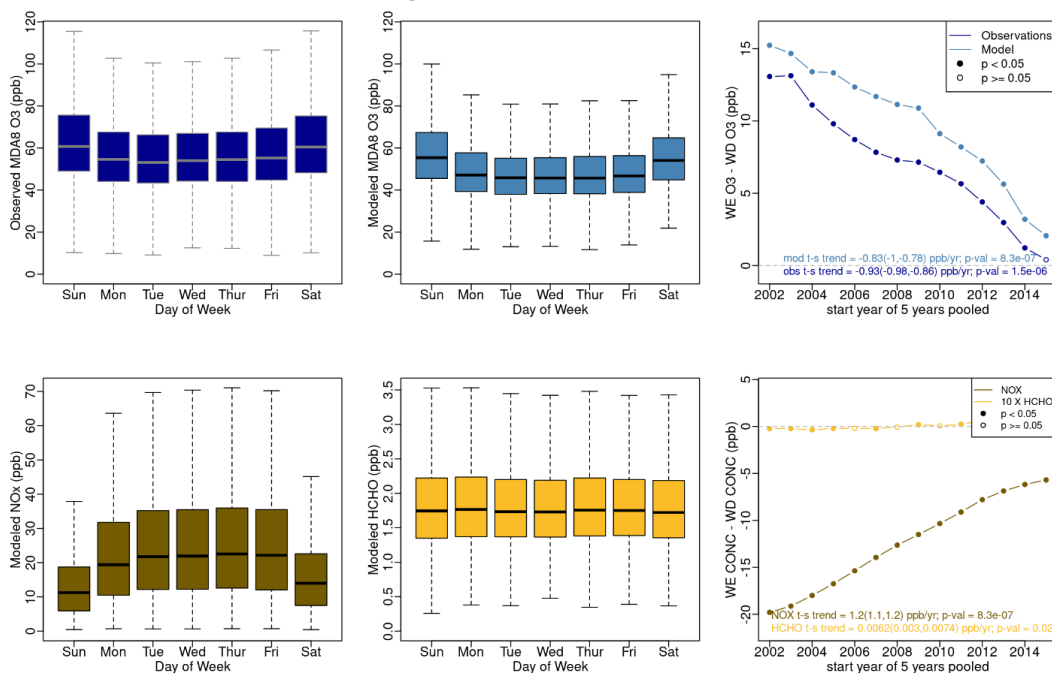


206

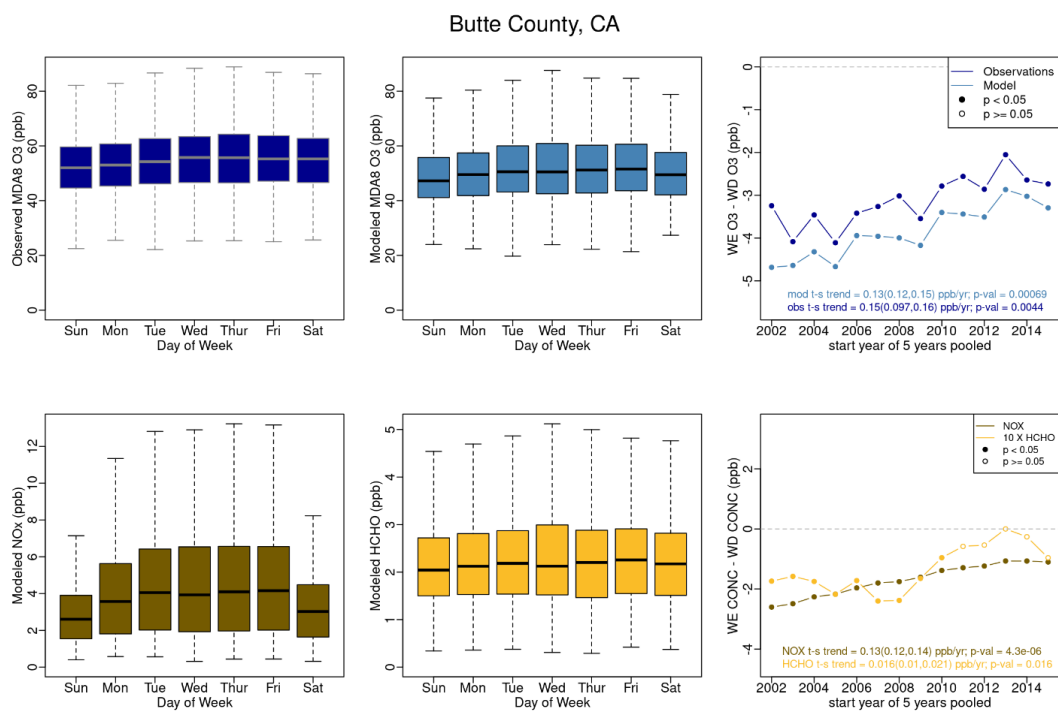


207 **Figure 1. Denver area 2002-2019 May-Sep: observed (top left) and modeled (top center) MDA8 ozone distribution by day**
 208 **of week; modeled NO_x (bottom left) and modeled formaldehyde (bottom center) distribution by day of week; observed and**
 209 **modeled trends in $\Delta\bar{O}_{3,DOW}$ (top right); modeled trends in WE-WD NO_x and formaldehyde differences (bottom right). The**
 210 **distributions by day of the week are for the entire 18 years with each box representing the 25th to 75th percentile for that**
 211 **day of the week across all 18 years, the whiskers representing the 1.5 times the interquartile range, and the bold line**
 212 **inside the box representing the median. WE-WD differences (top and bottom right) are based on 5-year rolling periods.**

Los Angeles-South Coast Air Basin, CA



213
 214 **Figure 2. Los Angeles area 2002-2019 May-Sep: observed (top left) and modeled (top center) MDA8 ozone distribution by**
 215 **day of week; modeled NO_x (bottom left) and modeled formaldehyde (bottom center) distribution by day of week; observed**
 216 **and modeled trends in $\Delta\bar{O}_{3,DOW}$ (top right); modeled trends in WE-WD NO_x and formaldehyde differences (bottom right).**
 217 **The distributions by day of the week are for the entire 18 years with each box representing the 25th to 75th percentile for**
 218 **that day of the week across all 18 years, the whiskers representing the 1.5 times the interquartile range, and the bold**
 219 **line inside the box representing the median. WE-WD differences (top and bottom right) are based on 5-year rolling periods.**



220

221 **Figure 3. Butte County, CA area 2002-2019 May-Sep: observed (top left) and modeled (top center) MDA8 ozone distribution**
 222 **by day of week; modeled NO_x (bottom left) and modeled formaldehyde (bottom center) distribution by day of week;**
 223 **observed and modeled trends in $\Delta O_{3,DOW}$ (top right); modeled trends in WE-WD NO_x and formaldehyde differences**
 224 **(bottom right). The distributions by day of the week are for the entire 18 years with each box representing the 25th to 75th**
 225 **percentile for that day of the week across all 18 years, the whiskers representing the 1.5 times the interquartile range, and**
 226 **the bold line inside the box representing the median. WE-WD differences (top and bottom right) are based on 5-year rolling**
 227 **periods.**

228

229 **3.2 Trend types of ozone day-of-week patterns**

230

231 Within any 5-year window, NO_x-saturated areas display a “weekend effect” meaning that ozone concentrations were
 232 statistically higher on weekends than on weekdays and NO_x-limited areas display a “weekday effect” meaning that
 233 ozone concentrations were statistically higher on weekdays than on weekends. We categorize the trends in ozone
 234 DOW patterns into 3 discrete categories: 1) disappearing weekend effect (i.e. areas that went from NO_x-saturated to
 235 NO_x-limited), 2) disappearing weekday effect (i.e. areas that went from NO_x-limited to approaching zero in terms of
 236 DOW differences), and 3) areas with no significant change over the 18-year time period. Disappearing weekend effect
 237 areas are characterized by a negative Thiel-Sen slope (e.g. Denver and Los Angeles in Figures 1 and 2 respectively).
 238 Disappearing weekday effect areas are characterized by a positive Thiel-Sen slope (e.g. Butte County in Figure 3).
 239 Areas with no trend are characterized by a lack of significance, as determined by the Mann-Kendall test. Trend types
 240 for all 51 areas based on observed and modeled datasets are shown in Figure 4 and 5.

241



242
243
244
245
246

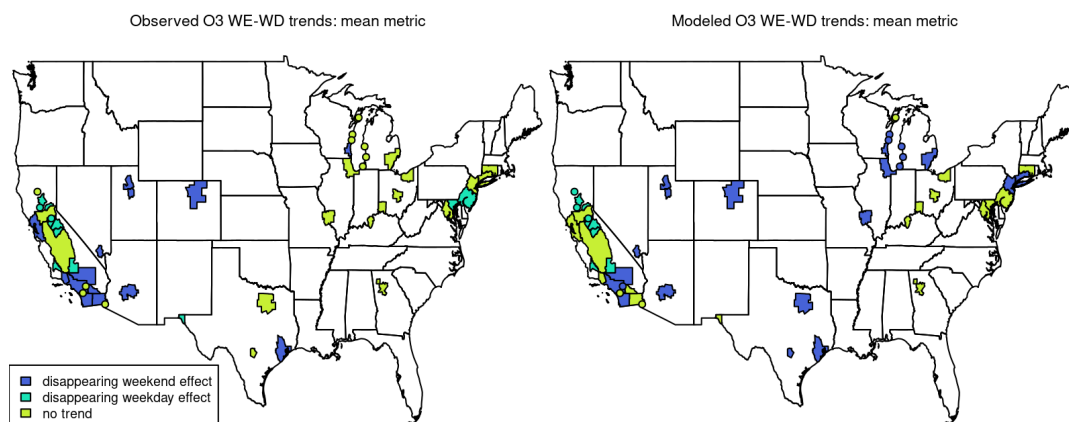


Figure 4. Map of ozone nonattainment areas color coded by trends in mean ozone day of week differences ($\Delta\overline{O_{3,DOW}}$) using observed data (left) and modeled data (right) over an 18-year period from 2002-2019. Ozone nonattainment areas less than 3000 km² in area are shown as dots on the map for visibility.

247
248
249
250
251

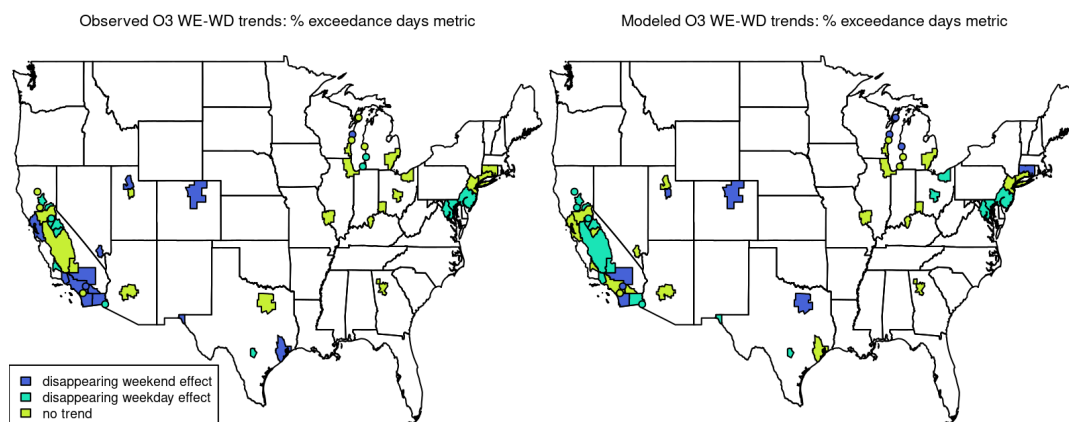


Figure 5. Map of ozone nonattainment areas color coded by trends in ozone day of week differences based on the percentage of days >70 ppb MDA8 ($\Delta O_{3,DOW,\%>70}$) using observed data (left) and modeled data (right) over an 18-year period from 2002-2019. Ozone nonattainment areas less than 3000 km² in area are shown as dots on the map for visibility.

252 3.2.1 “Disappearing weekend effect” case studies

253
254
255
256
257
258
259

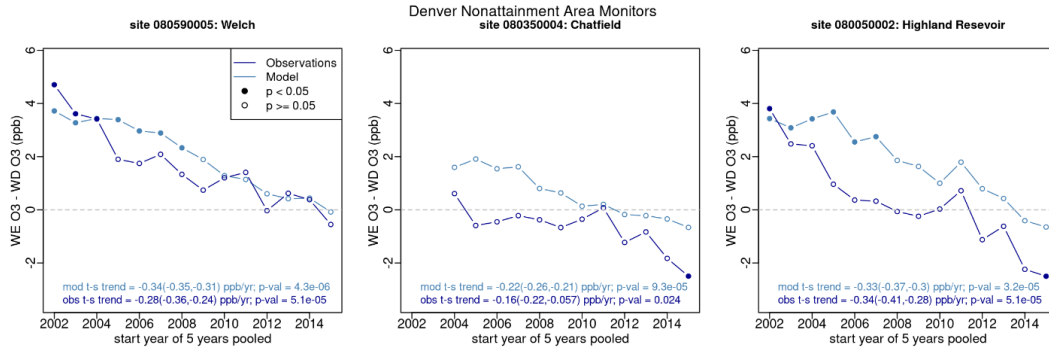
The disappearing weekend effect trend is typical of areas that initially had strongly positive ozone WE-WD differences (i.e., mean MDA8 ozone is higher on weekends than on weekdays), suggesting NO_x-saturated conditions, at the beginning of the analysis period. These areas typically transition into significant negative WE-WD MDA8 O₃ differences by the most recent 5-year window, suggesting a shift to NO_x-limited conditions by the end of the analysis period. Of the 51 nonattainment areas analyzed, 15 exhibit this type of trend based on observed data and 23 based on modeled data for $\Delta\overline{O_{3,DOW}}$. Two areas that exhibit this trend for $\Delta\overline{O_{3,DOW}}$ are Denver and Los Angeles shown in



260 Figures 1 and 2 respectively. In Denver, the modeled and observed $\overline{\Delta O_{3,DOW}}$ are statistically significant and in the
261 range of +3 to +4 ppb at the beginning of the analysis period. Both the model and observed data have statistically
262 significant decreasing Thiel-Sen slopes for $\overline{\Delta O_{3,DOW}}$, -0.29 ppb/yr and -0.23 ppb/yr for Denver and Los Angeles
263 respectively. In the most recent 2015-2019 5-year window, both modeled and observed $\overline{\Delta O_{3,DOW}}$ are negative and
264 statistically different from zero, suggesting a shift to NO_x-limited conditions. While the results shown in Figure 1
265 represent aggregated measured ozone data across all Denver nonattainment area monitors, Figure 6 shows behavior
266 at three specific monitors in Denver with monitoring records covering the majority of the analysis period. All three
267 sites were located to the south and southwest of the Denver urban area. The Welch monitor is located closer to the
268 Denver urban area in proximity to two major highways. While the monitored and modeled negative Thiel-Sen slopes
269 for $\overline{\Delta O_{3,DOW}}$ holds at all 3 sites, there are differences in the magnitude of the slopes and the sign of $\overline{\Delta O_{3,DOW}}$ across
270 sites. For instance, the Welch and Highland Reservoir sites both have statistically significant positive $\overline{\Delta O_{3,DOW}}$ at the
271 beginning of the analysis period suggesting both sites were NO_x-saturated in the early 2000s. While the Chatfield site
272 had positive $\overline{\Delta O_{3,DOW}}$ at the beginning of the analysis period, the differences were not statistically different from zero,
273 suggesting that this location may have already been transitioning to NO_x-limited conditions in the early-to-mid 2000s.
274 The model predicts that all three sites have non-significant negative $\overline{\Delta O_{3,DOW}}$ at the end of the analysis period while
275 observations show the negative $\overline{\Delta O_{3,DOW}}$ to be statistically significant at Chatfield and Highland Reservoir. This
276 suggests that the model may understate the NO_x-limited conditions in recent years at these locations. Los Angeles
277 provides another example of an area where both the model and the observations had strongly positive $\overline{\Delta O_{3,DOW}}$ at the
278 beginning of the analysis period and disappearing weekend effect trends (Figure 2). Similar to Denver, site to site
279 differences in the magnitude of $\overline{\Delta O_{3,DOW}}$ are evident in Los Angeles (Figure S-7) but the disappearing weekend effect
280 trend is fairly consistent across sites. Similar types of trends in Chicago and Houston are shown in supplemental
281 figures S-2 and S-3.

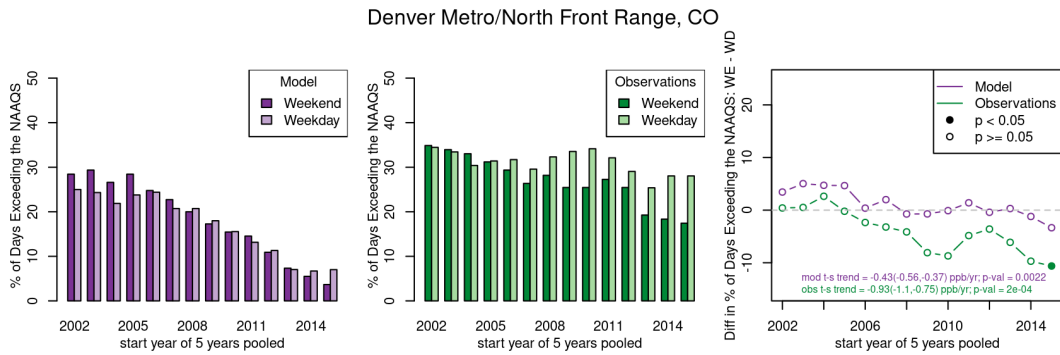
282
283 In general, similar disappearing weekend effect trends in $\Delta O_{3,DOW,\%>70}$ are evident, however this metric appears to be
284 noisier perhaps because it is capturing the frequency of extreme events which have a more stochastic nature than mean
285 ozone differences. Specifically, since there are a low number of exceedance days for most nonattainment areas in any
286 given year, a metric based on the percentage of those days falling on a Sunday versus a Tuesday, Wednesday or
287 Thursday will be inherently more noisy than a metric based on mean values. Figures 7 and 8 show $\Delta O_{3,DOW,\%>70}$
288 Thiel-Sen trends for Denver and Los Angeles. In both cases, the model underpredicts both the percentage of days with
289 MDA8 O₃ > 70 ppb and the Thiel-Sen slope. Additional examples of results for $\Delta O_{3,DOW,\%>70}$ are provided for
290 Chicago, Houston and New York City in Figure S-9, S-10 and S-11 respectively.

291
292



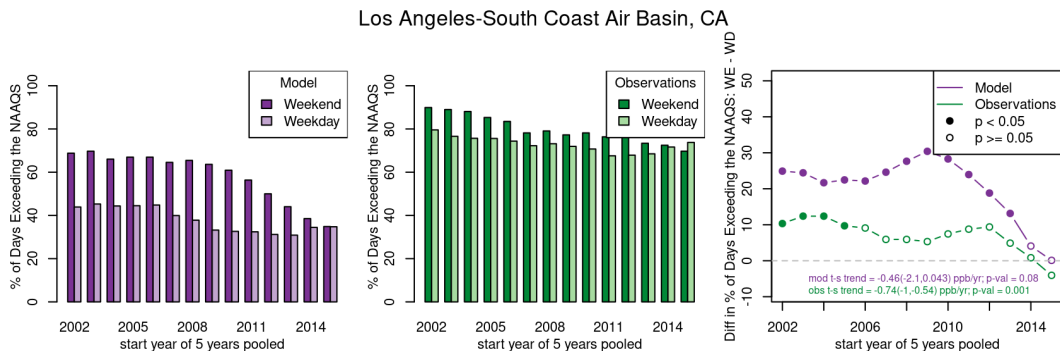
293
 294
 295
 296

Figure 6. Observed and modeled May-Sep trends in mean ozone day of week differences ($\overline{\Delta O_{3,DOW}}$) at three Denver area monitoring locations for 2002-2019 plotted as 5-year rolling periods.



297
 298
 299
 300
 301
 302
 303

Figure 7. Modeled (left) and observed (center) percent of days with MDA8 ozone exceeding 70 ppb at any monitor within the Denver nonattainment area during May-Sep on weekends and weekdays for 5-year rolling periods between 2002-2019; Observed and modeled trends in May-Sep $\Delta O_{3,DOW,\%>70}$ at Denver area monitors for 5-year rolling periods between 2002-2019 (right).



304
 305
 306
 307
 308
 309

Figure 8. Modeled (left) and observed (center) percent of days with MDA8 ozone exceeding 70 ppb at any monitor within the Los Angeles nonattainment area during May-Sep on weekends and weekdays for 5-year rolling periods between 2002-2019; Observed and modeled trends in May-Sep $\Delta O_{3,DOW,\%>70}$ at Los Angeles area monitors for 5-year rolling periods between 2002-2019 (right).

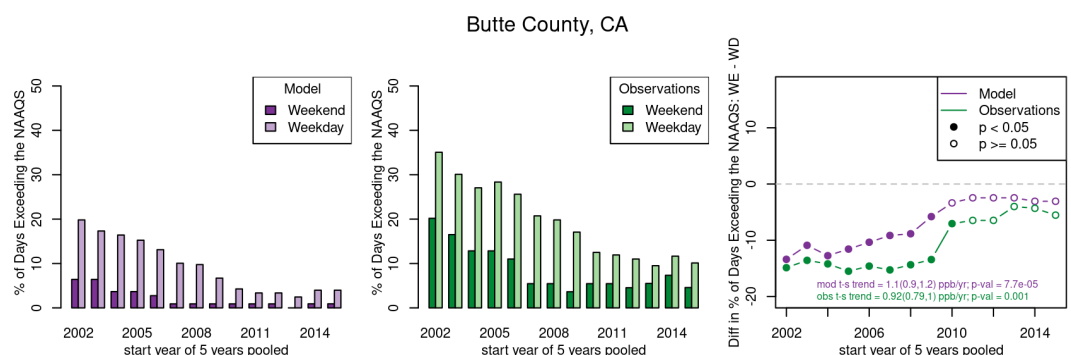


310 **3.2.2 “Disappearing weekday effect” case study**

311

312 The disappearing weekday effect trend type in $\overline{\Delta O_{3,DOW}}$ is evident in 12 out of the 51 nonattainment areas using
 313 observed data and 11 out of the 51 nonattainment areas using modeled data (Figure 4). This trend type is characterized
 314 by negative $\overline{\Delta O_{3,DOW}}$ values (i.e., weekday ozone higher than weekend ozone) throughout the analysis period
 315 indicating NO_x-limited conditions trending upwards toward zero which appears primarily in rural/agricultural areas
 316 in California. The Butte County nonattainment area in California is one example of an area exhibiting this type of day-
 317 of-week trend pattern as is evident using both $\overline{\Delta O_{3,DOW}}$ and $\Delta O_{3,DOW,\%>70}$ (Figures 3 and 9 respectively). The
 318 disappearing weekday effect could indicate that sources without day-of-week activity patterns are becoming more
 319 dominant contributors to local NO_x emissions. In that case, the day-of-week patterns for ambient NO_x concentrations
 320 are becoming less pronounced which would result in reductions in day-of-week ozone patterns. An alternate
 321 explanation is that local NO_x emissions in general have decreased substantially enough that local ozone formation has
 322 become less important in such areas and a larger fraction of total ozone is being transported from upwind sources. In
 323 that case, the origin of the transported ozone could be a mixture of multiple source areas that are at varying distances
 324 upwind which could lead to a loss in the day-of-week ozone signal. More analysis would be needed to investigate this
 325 idea with respect to nonattainment areas of interest.

326



327

328 **Figure 9. Modeled (left) and observed (center) percent of days with MDA8 ozone exceeding 70 ppb at any monitor within**
 329 **the Butte County, CA nonattainment area during May-Sep on weekends and weekdays for 5-year rolling periods between**
 330 **2002-2019; Observed and modeled trends in May-Sep $\Delta O_{3,DOW,\%>70}$ at Butte County, CA area monitors for 5-year rolling**
 331 **periods between 2002-2019 (right).**

332

333 **3.2.3 “No trend” case studies**

334

335 Out of the 51 nonattainment areas analyzed, 25 do not have a statistically significant $\overline{\Delta O_{3,DOW}}$ trend based on a p-value
 336 cut-off of 0.05 using observed data and 18 do not have a statistically significant trend using modeled data. The reason
 337 for the lack of trends may vary by area. Plots for several areas are provided in the supplemental information. Figures
 338 S-4, S-8 and S-11 provide the analysis for New York City which shows no trend for the $\overline{\Delta O_{3,DOW}}$ using observations
 339 but a statistically significant disappearing weekend effect trend for this metric using modeled data. Neither the model
 340 nor the observations show a significant trend in $\Delta O_{3,DOW,\%>70}$. One possible explanation for the lack of trends in New



341 York is the complex nature of the emissions sources and the meteorology impacting ozone formation in this area.
342 Figure S-8 shows $\Delta\overline{O_{3,DOW}}$ trends at three monitors in the New York City nonattainment area occurring in very
343 different locations. The Bronx IS 52 monitor, which is located in an urbanized part of the nonattainment area, shows
344 significant disappearing weekend effect in both modeled and observed $\Delta\overline{O_{3,DOW}}$. In contrast the Long Island –
345 Riverhead monitor and the Bridgeport CT monitor are both located in portions of the nonattainment area that are
346 typically downwind of the urban core on high ozone days and are impacted by complex meteorology associated with
347 the land-water interface near the Long Island sound. The modeled and observed data do not show significant
348 $\Delta\overline{O_{3,DOW}}$ trends at the Long Island site and only the model shows disappearing weekend effects trends at the CT site.
349 Due to the complex nature of this large urban area, some sites may not show trends at all and trends at other sites may
350 be masked when aggregating data across a large number of sites.

351
352 Several nonattainment areas appear to have negative slopes in $\Delta\overline{O_{3,DOW}}$ at the beginning of the analysis period and
353 positive slopes at the end of the analysis period resulting in no overall trend taken over the entire period. Cincinnati,
354 OH-KY exemplifies this pattern and on closer inspection the patterns appear to mirror annual changes in WE-WD
355 patterns in multiple meteorological parameters (Figure S-12). For Cincinnati the correlation coefficients between WE-
356 WD MDA8 O₃ differences and WE-WD meteorological parameter differences were 0.77, -0.83, 0.79, 0.89, -0.94, and
357 -0.73 for daily maximum temperature, daily average relative humidity, daily maximum planetary boundary layer
358 height, solar radiation, percent cloud cover and 24-hour transport direction respectively. Other areas exhibiting this
359 behavior are all located in relatively close proximity to Cincinnati, including Louisville, KY-IN and St. Louis, MO-
360 IL and to a lesser extent Columbus, OH and Atlanta, GA. These findings suggest that for these areas even five-year
361 processing blocks may not be sufficient to remove the effects of spurious weekly meteorological variations on ozone.
362 Figure S-13 shows that the correlation between WE-WD differences in seven meteorological variables and observed
363 $\Delta\overline{O_{3,DOW}}$ do not appear to be a driving factor in significant $\Delta\overline{O_{3,DOW}}$ trends in other areas but it is possible that some
364 additional areas which do not have statistically significant trends in $\Delta\overline{O_{3,DOW}}$ may also be impacted by meteorological
365 variations.

366

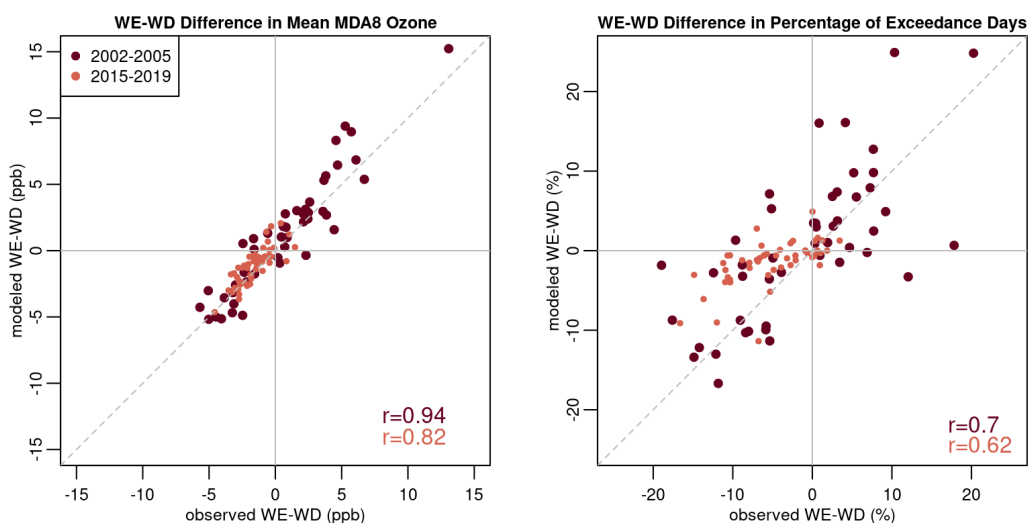
367 **3.3 Comparison of modeled and observed trends in ozone day-of-week patterns**

368

369 The modeled and observed trends in WE-WD differences for each of the 51 nonattainment areas are provided in
370 supplemental tables S1 ($\Delta\overline{O_{3,DOW}}$) and S2 ($\Delta O_{3,DOW,\%>70}$). Figure 10 provides a comparison of modeled to observed
371 WE-WD differences across the 51 nonattainment areas at the beginning of the analysis period (2002-2006) and at the
372 end of the analysis period (2015-2019). Each point represents the WE-WD MDA8 ozone difference for a single
373 nonattainment area, with the left-hand panel showing $\Delta\overline{O_{3,DOW}}$ and the right-hand panel showing $\Delta O_{3,DOW,\%>70}$. Data
374 points falling in the upper right quadrant of each panel represent areas for which both the observations and the modeled
375 DOW patterns suggest NO_x-saturated conditions. Data points in the lower left quadrant of each panel represent areas
376 for which both the observations and the model DOW patterns suggest NO_x-limited conditions. In the earlier 2002-
377 2006 time period, there are a large number of areas falling in both the upper right and lower left quadrants for both



378 metrics. In the 2015-2019 time period, almost all areas are located in the lower left quadrant for both metrics
379 suggesting that most US nonattainment areas have transitioned into NO_x-limited conditions. The correlation of
380 modeled and observed WE-WD differences is quite high ($r = 0.94$ and 0.82 for $\overline{\Delta O_{3,DOW}}$ in the earliest and most recent
381 time periods, respectively, and $r = 0.7$ and 0.62 for $\Delta O_{3,DOW,\%>70}$ in the earliest and most recent time periods,
382 respectively). For both metrics, the majority of points fall above the 1:1 line indicating that, in general, the model
383 overestimated the degree of NO_x-saturated conditions and underestimated the degree of NO_x-limited conditions.
384



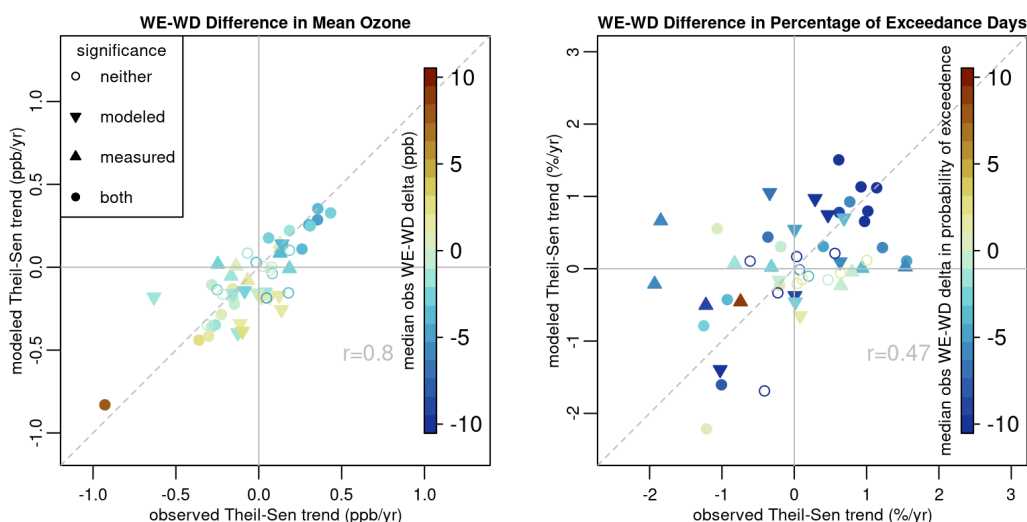
385
386 **Figure 10. Comparison of modeled and observed WE-WD MDA8 O₃ differences for $\overline{\Delta O_{3,DOW}}$ (left panel) and $\Delta O_{3,DOW,\%>70}$**
387 **(right panel). Differences shown for the 2002-2006 time period and for the 2015-2019 time period. Each dot represents a**
388 **different nonattainment area.**
389

390 Maps in Figures 4 and 5 show the locations of areas predicted to have disappearing weekend effect trends, disappearing
391 weekday effect trends and no trends for $\overline{\Delta O_{3,DOW}}$ and $\Delta O_{3,DOW,\%>70}$ respectively. The maps show general consistency
392 among which areas are predicted to have each trend type between observations and the model, although some areas
393 predicted to have significant trends with one dataset or with one metric do not have significant trends with the other
394 dataset or metric. Nine areas are predicted to have disappearing weekend effect trends in both datasets and with both
395 metrics indicating strong agreement that they are shifting to more NO_x-limited conditions: Milwaukee, WI; Houston,
396 TX; Phoenix, AZ; Denver, CO; Northern Wasatch Front, UT; Southern Wasatch Front, UT; Las Vegas, NV; Los
397 Angeles – San Bernardino County, CA; Los Angeles – South Coast, CA; and San Diego, CA.
398

399 Figure 11 compares modeled and observed Thiel-Sen slopes in WE-WD MDA8 O₃ differences across all areas. Each
400 point represents a single nonattainment area color-coded by median $\overline{\Delta O_{3,DOW}}$ or $\Delta O_{3,DOW,\%>70}$. The correlation of
401 modeled versus observed Thiel-Sen slopes using $\overline{\Delta O_{3,DOW}}$ is stronger ($r = 0.8$) than the correlation using $\Delta O_{3,DOW,\%>70}$
402 ($r = 0.47$). While the model does not always correctly predict the Thiel-Sen slope, the data falls close to the 1:1 line
403 for the $\overline{\Delta O_{3,DOW}}$ suggesting that the model does not systematically over or under predict the trends in WE-WD

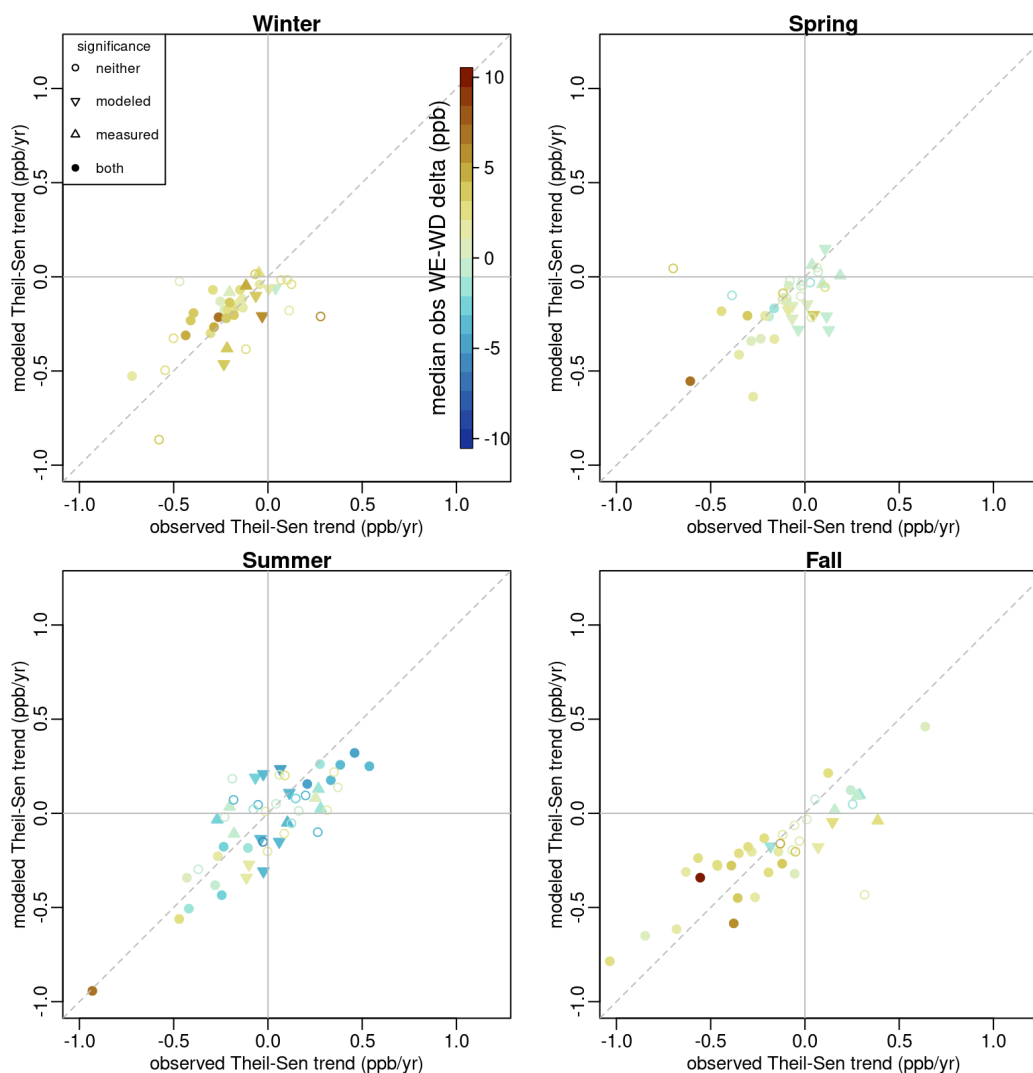


404 differences from 2002-2019. The trend types described above for $\overline{\Delta O_{3,DOW}}$ metric are visible in the left-panel of Figure
405 11. Most NO_x-saturated areas (yellow and brown symbols) and some NO_x-limited areas (blue symbols) have negative
406 Thiel-Sen slopes towards NO_x-limited conditions similar to those described above for Denver and Los Angeles
407 (shown as the dark brown symbol at the bottom-left of the plot). Areas with positive Thiel-Sen slopes tend to be the
408 most NO_x-limited areas (darker blue symbols) and represent the disappearing weekday trends demonstrated by Butte
409 County. The model is not as accurate at predicting $\Delta O_{3,DOW,\%>70}$ Thiel-Sen slopes as $\overline{\Delta O_{3,DOW}}$ Thiel-Sen slopes, as
410 evidenced by the increased scatter in the right-hand panel of Figure 11 compared to the left-hand panel. Some areas
411 have few exceedances of the NAAQS in the later years of the trends period and this small sample size could explain
412 the difference between the monitored and modeled slopes, given that the model predicted fewer exceedance days than
413 were observed in many areas.
414



415
416 **Figure 11. Comparison of modeled and observed Thiel-Sen slopes in May-Sep WE-WD MDA8 O₃ differences across all**
417 **nonattainment areas for $\overline{\Delta O_{3,DOW}}$ (left panel) and $\Delta O_{3,DOW,\%>70}$ (right panel). Whether or not the trend is significant with**
418 **observed and/or modeled data is indicated by the shape of the symbol. Median WE-WD differences across all years are**
419 **indicated by the color scale with positive differences (NO_x-saturated areas) shown in shades of yellow and brown and**
420 **negative differences (NO_x-limited areas) shown in shades of blue. Note that the brown symbol on both figures represents**
421 **the Los Angeles nonattainment area.**
422

423 Figure 12 shows the comparison of $\overline{\Delta O_{3,DOW}}$ Thiel-Sen slopes by season. The summer plot looks similar to the May-
424 September plot shown in Figure 11. Winter, spring, and fall data show median $\overline{\Delta O_{3,DOW}}$ near zero or greater than zero
425 in most nonattainment areas suggesting transitional or NO_x-saturated conditions in these seasons. Both observations
426 and model predictions suggest $\overline{\Delta O_{3,DOW}}$ negative Thiel-Sen slopes in these seasons suggesting that nonattainment
427 areas in the US may be transitioning towards NO_x-limited conditions even outside of the summer ozone season.



428

429 **Figure 12.** Comparison of modeled and observed $\Delta O_{3,DOW}$ Thiel-Sen slopes across all nonattainment areas in winter (top
430 left), spring (top right), summer (bottom left) and fall (bottom right). Whether or not the trend is significant with observed
431 and/or modeled data is indicated by the shape of the symbol. Median WE-WD differences across all years are indicated by
432 the color scale with positive differences (NO_x-saturated areas) shown in shades of yellow and brown and negative
433 differences (NO_x-limited areas) shown in shades of blue. Note that year-round ozone monitoring is not required in some
434 parts of the US and therefore monitoring data may not be available outside the May-September period in some areas.

435 4 Conclusions

436

437 While this assessment has provided insight into the ozone formation regimes across high-ozone locations in the US,
438 some key questions remain about the important drivers for year-to-year changes in DOW ozone patterns and which
439 of those drivers are well captured by the EQUATES dataset. First, while NO_x and VOC emissions have been steadily
440 decreasing across most areas of the US, exceptions to that pattern include increasing wildfire emissions especially in



441 the Western US and increasing emissions from oil and gas activities near US nonattainment areas in Texas, Colorado,
442 New Mexico and Utah. Future work could focus on areas impacted by these two emissions sources to assess both the
443 impact of these increasing emissions on ozone formation regimes and the ability of the EQUATES dataset to capture
444 those impacts. Second, this assessment predominantly focused on ozone values across the May-Sep ozone season,
445 however, past work has identified some seasonally varying ozone biases within the CMAQ model (Appel et al., 2021).
446 Specifically, EQUATES has a tendency to underpredict ozone during the spring and overpredict ozone later in the
447 summer (Figures S-14 and S-15). Given that ozone formation tends to be more NO_x-saturated in the springtime than
448 in the summer (Jin et al., 2020; Jin et al., 2017), a more in-depth assessment would be needed to fully characterize the
449 extent that differences in observed and modeled WE-WD ozone differences are impacted by this seasonally varying
450 model performance. Third, we assessed DOW ozone patterns across multiple complex urban areas that encompassed
451 spatially heterogeneous emissions sources and meteorology. For some of these areas (e.g. Los Angeles, CA and
452 Denver, CO) the sign of the Thiel-Sen slopes in WE-WD ozone appeared consistent across monitoring locations while
453 in others (e.g. New York City, NY) different monitoring locations across the area appeared to show different types of
454 trends. Further local scale investigation into each of these areas would be necessary to fully characterize the nuances
455 of DOW and year-to-year variations in emission and meteorology that obscure the ozone DOW trends in some areas
456 but not others when aggregating across monitor locations in those areas. Finally, an intriguing trend in ozone DOW
457 patterns was identified in multiple rural and agricultural areas of California. Recent literature has suggested that soil
458 NO emissions, which are unlikely to have a DOW emissions pattern, are an important NO_x emissions source in
459 agricultural locations of California (Almaraz et al., 2018; Zhu et al., 2023). Could the ozone DOW trends observed in
460 these areas be reflective of the increasing relative importance of NO_x sources other than mobile sources in those
461 locations? More assessment is needed to definitively determine whether the trend in a decreasing weekday effect is a
462 reliable indicator of areas that are becoming more dominated by local NO_x sources that do not vary by DOW, more
463 dominated by transported ozone, or some other factor. It is important to note that transported ozone may come from
464 nearby regional sources or from longer range sources provided the transport times are sufficient to mask any DOW
465 patterns that would be evident in the source region.

466

467 In this analysis we found that trends in ozone formation chemistry may not always be clearly shown by trends in DOW
468 patterns which are impacted by a complex set of local factors including meteorology, the mix of local emissions
469 sources and monitor locations in relationship to land-water interfaces. Lack of trends appear more often using observed
470 data than modeled data (Figures 4 and 5) meaning that, while the model accurately captures Thiel-Sen slopes for
471 $\overline{\Delta O_{3,DOW}}$ and $\Delta O_{3,DOW,\%>70}$ (Figure 11), p-values below 0.05 are less common using observational data. This suggests
472 that there may be some stochastic processes making observed year-to-year WE-WD ozone differences noisy which
473 are not fully captured by the model. Even with these limitations, this analysis has shown that DOW patterns in ambient
474 NO_x concentrations persist in US urban areas but have become less prominent over the 18-year period analyzed. These
475 DOW NO_x differences have resulted in distinctive DOW ozone patterns in many of the nonattainment areas assessed.
476 The EQUATES modeling simulations appear to show larger and more positive WE-WD ozone differences than
477 observational data suggesting that ozone formation in this modeling dataset is less NO_x-limited than in the



478 observations. Despite this discrepancy, the EQUATES dataset captures year-to-year changes in WE-WD ozone
479 patterns as demonstrated by high correlation of the Thiel-Sen slopes for WE-WD ozone differences. Both the WE-
480 WD ozone trends and agreement between the modeled and observation datasets are more apparent when assessing
481 summertime mean MDA8 ozone than when analyzing extreme values using the percentage of exceedance days metric.
482 Assessing frequencies or magnitudes of extreme values is challenging using a dataset with a limited number of
483 weekend and weekday days due to the stochastic and infrequent nature of high ozone events in many areas.

484

485 While there are multiple types of measurements and modeling assessments that can be applied to characterize local
486 ozone formation regimes, many of these require specialized measurements or datasets that are not readily available in
487 all areas. In contrast, assessing DOW ozone patterns requires only routine daily ozone measurements that are widely
488 available across urban areas in the US and in other countries. Consequently, this type of assessment is a useful tool
489 and may be applied in many areas using routine measurements. In locations with long-term measurements, DOW
490 patterns offer a method to look at trends in ozone formation chemistry over time. While DOW patterns in ozone are
491 especially useful given the wide availability of data required for this type of assessment, we anticipate that in the near
492 future additional datasets for assessing ozone chemical formation regimes will become more widely available.
493 Specifically, O₃, NO₂ and HCHO data from the recently launched TEMPO satellite may provide the ability to better
494 understand the relationships between WE-WD ozone patterns and precursor concentrations.

495 **Author contributions**

496 All authors contributed to conceptualization of the project. HS, CH, KF, BW, and WA contributed to data curation.
497 HS conducted formal analysis. HS, CH, AW, KF, BW, BH, and SK contributed to developing the methodology.
498 HS and BW developed software for performing the analysis. HS, CH, AW, JL, NP, BW, and GT contributed to
499 validation. HS, BW, and BH helped visualize the data. All authors contributed to the writing and editing of the
500 manuscript.

501 **Competing interests**

502 The authors declare that they have no conflict of interest.

503

504 *Disclaimer:* The views expressed in this manuscript are those of the authors and do not necessarily reflect the views
505 or policies of the U.S. Environmental Protection Agency.

506 **Acknowledgements**

507 The authors would like to acknowledge Chris Nolte and Golam Sarwar for helpful comments on this manuscript.

508 **References**

509 Adame, J. A., Hernández-Ceballos, M. Á., Sorribas, M., Lozano, A., and Morena, B. A. D. I.: Weekend-
510 Weekday Effect Assessment for O₃, NO_x, CO and PM₁₀ in Andalusia, Spain (2003-2008), *Aerosol and Air*
511 *Quality Research*, 14, 1862-1874, 10.4209/aaqr.2014.02.0026, 2014.
512 Almaraz, M., Bai, E., Wang, C., Trousdell, J., Conley, S., Faloon, I., and Houlton, B. Z.: Agriculture is a
513 major source of NO_x pollution in California, *Science Advances*, 4, eaao3477,
514 doi:10.1126/sciadv.aao3477, 2018.



515 Appel, K. W., Bash, J. O., Fahey, K. M., Foley, K. M., Gilliam, R. C., Hogrefe, C., Hutzell, W. T., Kang, D.,
516 Mathur, R., Murphy, B. N., Napelenok, S. L., Nolte, C. G., Pleim, J. E., Pouliot, G. A., Pye, H. O. T., Ran, L.,
517 Roselle, S. J., Sarwar, G., Schwede, D. B., Sidi, F. I., Spero, T. L., and Wong, D. C.: The Community
518 Multiscale Air Quality (CMAQ) model versions 5.3 and 5.3.1: system updates and evaluation, *Geosci.*
519 *Model Dev.*, 14, 2867-2897, 10.5194/gmd-14-2867-2021, 2021.

520 Atkinson-Palombo, C. M., Miller, J. A., and Balling, R. C.: Quantifying the ozone “weekend effect” at
521 various locations in Phoenix, Arizona, *Atmospheric Environment*, 40, 7644-7658,
522 <https://doi.org/10.1016/j.atmosenv.2006.05.023>, 2006.

523 Blanchard, C. L. and Tanenbaum, S.: Weekday/Weekend Differences in Ambient Air Pollutant
524 Concentrations in Atlanta and the Southeastern United States, *Journal of the Air & Waste Management*
525 *Association*, 56, 271-284, 10.1080/10473289.2006.10464455, 2006.

526 Blanchard, C. L., Tanenbaum, S., and Lawson, D. R.: Differences between Weekday and Weekend Air
527 Pollutant Levels in Atlanta; Baltimore; Chicago; Dallas–Fort Worth; Denver; Houston; New York; Phoenix;
528 Washington, DC; and Surrounding Areas, *Journal of the Air & Waste Management Association*, 58, 1598-
529 1615, 10.3155/1047-3289.58.12.1598, 2008.

530 Bruntz, S. M., Cleveland, W. S., Graedel, T. E., Kleiner, B., and Warner, J. L.: OZONE CONCENTRATIONS IN
531 NEW-JERSEY AND NEW-YORK - STATISTICAL ASSOCIATION WITH RELATED VARIABLES, *Science*, 186, 257-
532 259, 10.1126/science.186.4160.257, 1974.

533 Chinkin, L. R., Coe, D. L., Funk, T. H., Hafner, H. R., Roberts, P. T., Ryan, P. A., and Lawson, D. R.: Weekday
534 versus Weekend Activity Patterns for Ozone Precursor Emissions in California’s South Coast Air Basin,
535 *Journal of the Air & Waste Management Association*, 53, 829-843, 10.1080/10473289.2003.10466223,
536 2003.

537 Cleveland, W. S., Graedel, T. E., Kleiner, B., and Warner, J. L.: SUNDAY AND WORKDAY VARIATIONS IN
538 PHOTOCHEMICAL AIR-POLLUTANTS IN NEW-JERSEY AND NEW-YORK, *Science*, 186, 1037-1038,
539 10.1126/science.186.4168.1037, 1974.

540 de Foy, B., Brune, W. H., and Schauer, J. J.: Changes in ozone photochemical regime in Fresno, California
541 from 1994 to 2018 deduced from changes in the weekend effect, *Environmental Pollution*, 263, 114380,
542 <https://doi.org/10.1016/j.envpol.2020.114380>, 2020.

543 EPA, U.: EQUATESv1.0: Emissions, WRF/MCIP, CMAQv5.3.2 Data -- 2002-2019 US_12km and
544 NHEMI_108km (V5), UNC Dataverse [dataset], doi:10.15139/S3/F2KJJK, 2021.

545 Fisher, R. A.: The Logic of Inductive Inference, *Journal of the Royal Statistical Society*, 98, 39-82,
546 10.2307/2342435, 1935.

547 Foley, K. M., Pouliot, G. A., Eyth, A., Aldridge, M. F., Allen, C., Appel, K. W., Bash, J. O., Beardsley, M.,
548 Beidler, J., Choi, D., Farkas, C., Gilliam, R. C., Godfrey, J., Henderson, B. H., Hogrefe, C., Koplitz, S. N.,
549 Mason, R., Mathur, R., Misenis, C., Possiel, N., Pye, H. O. T., Reynolds, L., Roark, M., Roberts, S.,
550 Schwede, D. B., Seltzer, K. M., Sonntag, D., Talgo, K., Toro, C., Vukovich, J., Xing, J., and Adams, E.: 2002–
551 2017 anthropogenic emissions data for air quality modeling over the United States, *Data in Brief*, 47,
552 109022, <https://doi.org/10.1016/j.dib.2023.109022>, 2023.

553 Fujita, E. M., Stockwell, W. R., Campbell, D. E., Keislar, R. E., and Lawson, D. R.: Evolution of the
554 Magnitude and Spatial Extent of the Weekend Ozone Effect in California’s South Coast Air Basin, 1981–
555 2000, *Journal of the Air & Waste Management Association*, 53, 802-815,
556 10.1080/10473289.2003.10466225, 2003a.

557 Fujita, E. M., Campbell, D. E., Zielinska, B., Sagebiel, J. C., Bowen, J. L., Goliff, W. S., Stockwell, W. R., and
558 Lawson, D. R.: Diurnal and Weekday Variations in the Source Contributions of Ozone Precursors in
559 California’s South Coast Air Basin, *Journal of the Air & Waste Management Association*, 53, 844-863,
560 10.1080/10473289.2003.10466226, 2003b.



- 561 Gao, H. O.: Day of week effects on diurnal ozone/NO_x cycles and transportation emissions in Southern
562 California, *Transportation Research Part D: Transport and Environment*, 12, 292-305,
563 <https://doi.org/10.1016/j.trd.2007.03.004>, 2007.
- 564 Gao, H. O. and Niemeier, D. A.: The impact of rush hour traffic and mix on the ozone weekend effect in
565 southern California, *Transportation Research Part D: Transport and Environment*, 12, 83-98,
566 <https://doi.org/10.1016/j.trd.2006.12.001>, 2007.
- 567 Jaffe, D. A., Ninneman, M., and Chan, H. C.: NO_x and O₃ Trends at U.S. Non-Attainment Areas for 1995–
568 2020: Influence of COVID-19 Reductions and Wildland Fires on Policy-Relevant Concentrations, 127,
569 e2021JD036385, <https://doi.org/10.1029/2021JD036385>, 2022.
- 570 Jiménez, P., Parra, R., Gassó, S., and Baldasano, J. M.: Modeling the ozone weekend effect in very
571 complex terrains: a case study in the Northeastern Iberian Peninsula, *Atmospheric Environment*, 39,
572 429-444, <https://doi.org/10.1016/j.atmosenv.2004.09.065>, 2005.
- 573 Jin, X., Fiore, A., Boersma, K. F., Smedt, I. D., and Valin, L.: Inferring Changes in Summertime Surface
574 Ozone–NO_x–VOC Chemistry over U.S. Urban Areas from Two Decades of Satellite and Ground-Based
575 Observations, *Environmental Science & Technology*, 54, 6518-6529, [10.1021/acs.est.9b07785](https://doi.org/10.1021/acs.est.9b07785), 2020.
- 576 Jin, X., Fiore, A. M., Murray, L. T., Valin, L. C., Lamsal, L. N., Duncan, B., Folkert Boersma, K., De Smedt, I.,
577 Abad, G. G., Chance, K., and Tonnesen, G. S.: Evaluating a Space-Based Indicator of Surface Ozone-NO_x-
578 VOC Sensitivity Over Midlatitude Source Regions and Application to Decadal Trends, *Journal of*
579 *Geophysical Research: Atmospheres*, 122, 10,439-410,461, <https://doi.org/10.1002/2017JD026720>,
580 2017.
- 581 Kendall, M. G.: Rank Correlation Methods, 4th edition, Charles Griffin, London1975.
- 582 Koo, B., Jung, J., Pollack, A. K., Lindhjem, C., Jimenez, M., and Yarwood, G.: Impact of meteorology and
583 anthropogenic emissions on the local and regional ozone weekend effect in Midwestern US,
584 *Atmospheric Environment*, 57, 13-21, <https://doi.org/10.1016/j.atmosenv.2012.04.043>, 2012.
- 585 Koplitz, S., Simon, H., Henderson, B., Liljegren, J., Tonnesen, G., Whitehill, A., and Wells, B.: Changes in
586 Ozone Chemical Sensitivity in the United States from 2007 to 2016, *ACS Environmental Au*, 2, 206-222,
587 [10.1021/acsenvironau.1c00029](https://doi.org/10.1021/acsenvironau.1c00029), 2022.
- 588 Krotkov, N. A., McLinden, C. A., Li, C., Lamsal, L. N., Celarier, E. A., Marchenko, S. V., Swartz, W. H.,
589 Bucsel, E. J., Joiner, J., Duncan, B. N., Boersma, K. F., Veefkind, J. P., Levelt, P. F., Fioletov, V. E.,
590 Dickerson, R. R., He, H., Lu, Z. F., and Streets, D. G.: Aura OMI observations of regional SO₂ and NO₂
591 pollution changes from 2005 to 2015, *Atmospheric Chemistry and Physics*, 16, 4605-4629, [10.5194/acp-
592 16-4605-2016](https://doi.org/10.5194/acp-16-4605-2016), 2016.
- 593 Lamsal, L. N., Duncan, B. N., Yoshida, Y., Krotkov, N. A., Pickering, K. E., Streets, D. G., and Lu, Z. F.: U.S.
594 NO₂ trends (2005-2013): EPA Air Quality System (AQS) data versus improved observations from the
595 Ozone Monitoring Instrument (OMI), *Atmospheric Environment*, 110, 130-143,
596 [10.1016/j.atmosenv.2015.03.055](https://doi.org/10.1016/j.atmosenv.2015.03.055), 2015.
- 597 Mann, H. B.: Nonparametric Tests Against Trend, *Econometrica*, 13, 245-259, [10.2307/1907187](https://doi.org/10.2307/1907187), 1945.
- 598 Marr, L. C. and Harley, R. A.: Spectral analysis of weekday-weekend differences in ambient ozone,
599 nitrogen oxide, and non-methane hydrocarbon time series in California, *Atmospheric Environment*, 36,
600 2327-2335, [10.1016/s1352-2310\(02\)00188-7](https://doi.org/10.1016/s1352-2310(02)00188-7), 2002a.
- 601 Marr, L. C. and Harley, R. A.: Modeling the effect of weekday-weekend differences in motor vehicle
602 emissions on photochemical air pollution in central California, *Environmental Science & Technology*, 36,
603 4099-4106, [10.1021/es020629x](https://doi.org/10.1021/es020629x), 2002b.
- 604 Martins, E. M., Nunes, A. C. L., and Correa, S. M.: Understanding Ozone Concentrations During
605 Weekdays and Weekends in the Urban Area of the City of Rio de Janeiro, *Journal of the Brazilian*
606 *Chemical Society*, 26, 1967-1975, [10.5935/0103-5053.20150175](https://doi.org/10.5935/0103-5053.20150175), 2015.
- 607 Mehta, C. R. and Patel, N. R.: A Network Algorithm for Performing Fisher's Exact Test in $r \times c$ Contingency
608 Tables, *Journal of the American Statistical Association*, 78, 427-434, [10.2307/2288652](https://doi.org/10.2307/2288652), 1983.



- 609 Murphy, J. G., Day, D. A., Cleary, P. A., Wooldridge, P. J., Millet, D. B., Goldstein, A. H., and Cohen, R. C.:
610 The weekend effect within and downwind of Sacramento – Part 1: Observations of ozone,
611 nitrogen oxides, and VOC reactivity, *Atmos. Chem. Phys.*, 7, 5327-5339, 10.5194/acp-7-5327-2007, 2007.
612 Paschalidou, A. K. and Kassomenos, P. A.: Comparison of Air Pollutant Concentrations between
613 Weekdays and Weekends in Athens, Greece for Various Meteorological Conditions, *Environmental*
614 *Technology*, 25, 1241-1255, 10.1080/09593332508618372, 2004.
615 Pierce, T., Hogrefe, C., Trivikrama Rao, S., Porter, P. S., and Ku, J.-Y.: Dynamic evaluation of a regional air
616 quality model: Assessing the emissions-induced weekly ozone cycle, *Atmospheric Environment*, 44,
617 3583-3596, <https://doi.org/10.1016/j.atmosenv.2010.05.046>, 2010.
618 Pires, J. C. M.: Ozone Weekend Effect Analysis in Three European Urban Areas, *CLEAN – Soil, Air, Water*,
619 40, 790-797, <https://doi.org/10.1002/clen.201100410>, 2012.
620 Plocoste, T., Dorville, J.-F., Monjoly, S., Jacoby-Koaly, S., and André, M.: Assessment of nitrogen oxides
621 and ground-level ozone behavior in a dense air quality station network: Case study in the Lesser Antilles
622 Arc, *Journal of the Air & Waste Management Association*, 68, 1278-1300,
623 10.1080/10962247.2018.1471428, 2018.
624 Pun, B. K., Seigneur, C., and White, W.: Day-of-Week Behavior of Atmospheric Ozone in Three U.S. Cities,
625 *Journal of the Air & Waste Management Association*, 53, 789-801, 10.1080/10473289.2003.10466231,
626 2003.
627 Roberts, S. J., Salawitch, R. J., Wolfe, G. M., Marvin, M. R., Canty, T. P., Allen, D. J., Hall-Quinlan, D. L.,
628 Krask, D. J., and Dickerson, R. R.: Multidecadal trends in ozone chemistry in the Baltimore-Washington
629 Region, *Atmospheric Environment*, 285, 119239, <https://doi.org/10.1016/j.atmosenv.2022.119239>,
630 2022.
631 Rubio, M. A., Sanchez, K., and Lissi, Y. E.: OZONE LEVELS ASSOCIATED TO THE PHOTOCHEMICAL SMOG IN
632 SANTIAGO OF CHILE. THE ELUSIVE ROL OF HYDROCARBONS, *Journal of the Chilean Chemical Society*, 56,
633 709-711, 2011.
634 Russell, A. R., Valin, L. C., and Cohen, R. C.: Trends in OMI NO₂ observations over the United States:
635 effects of emission control technology and the economic recession, *Atmospheric Chemistry and Physics*,
636 12, 12197-12209, 10.5194/acp-12-12197-2012, 2012.
637 Seinfeld, J. H. and Pandis, S. N.: *Atmospheric chemistry and physics: from air pollution to climate change*,
638 John Wiley & Sons 2016.
639 Sen, P. K.: Estimates of the Regression Coefficient Based on Kendall's Tau, *Journal of the American*
640 *Statistical Association*, 63, 1379-1389, 10.1080/01621459.1968.10480934, 1968.
641 Sillman, S.: THE USE OF NO_y, H₂O₂, AND HNO₃ AS INDICATORS FOR OZONE-NO_x-HYDROCARBON
642 SENSITIVITY IN URBAN LOCATIONS, *Journal of Geophysical Research-Atmospheres*, 100, 14175-14188,
643 10.1029/94jd02953, 1995.
644 Sillman, S.: The relation between ozone, NO_x and hydrocarbons in urban and polluted rural
645 environments, *Atmospheric Environment*, 33, 1821-1845, 10.1016/s1352-2310(98)00345-8, 1999.
646 Sillman, S., Logan, J. A., and Wofsy, S. C.: THE SENSITIVITY OF OZONE TO NITROGEN-OXIDES AND
647 HYDROCARBONS IN REGIONAL OZONE EPISODES, *Journal of Geophysical Research-Atmospheres*, 95,
648 1837-1851, 10.1029/JD095iD02p01837, 1990.
649 Simon, H., Reff, A., Wells, B., Xing, J., and Frank, N.: Ozone Trends Across the United States over a Period
650 of Decreasing NO_x and VOC Emissions, *Environmental Science & Technology*, 49, 186-195,
651 10.1021/es504514z, 2015.
652 Singh, S. and Kavouras, I. G.: Trends of Ground-Level Ozone in New York City Area during
653 2007–2017, 13, 114, 2022.
654 Theil, H.: A Rank-Invariant Method of Linear and Polynomial Regression Analysis, in: Henri Theil's
655 Contributions to Economics and Econometrics: Econometric Theory and Methodology, edited by: Raj, B.,
656 and Koerts, J., Springer Netherlands, Dordrecht, 345-381, 10.1007/978-94-011-2546-8_20, 1992.



657 Toro, C., Foley, K., Simon, H., Henderson, B., Baker, K. R., Eyth, A., Timin, B., Appel, W., Luecken, D.,
658 Beardsley, M., Sonntag, D., Possiel, N., and Roberts, S.: Evaluation of 15 years of modeled atmospheric
659 oxidized nitrogen compounds across the contiguous United States, *Elementa-Science of the*
660 *Anthropocene*, 9, 10.1525/elementa.2020.00158, 2021.
661 U.S. Environmental Protection Agency: Integrated Science Assessment (ISA) for Particulate Matter (Final
662 report, Dec 2019). U.S. Environmental Protection Agency, Washington, DC, EPA/600/R-19/188, 2019.
663 Warneke, C., de Gouw, J. A., Edwards, P. M., Holloway, J. S., Gilman, J. B., Kuster, W. C., Graus, M., Atlas,
664 E., Blake, D., Gentner, D. R., Goldstein, A. H., Harley, R. A., Alvarez, S., Rappenglueck, B., Trainer, M., and
665 Parrish, D. D.: Photochemical aging of volatile organic compounds in the Los Angeles basin: Weekday-
666 weekend effect, 118, 5018-5028, <https://doi.org/10.1002/jgrd.50423>, 2013.
667 Welch, B. L.: THE GENERALIZATION OF 'STUDENT'S' PROBLEM WHEN SEVERAL DIFFERENT POPULATION
668 VARIANCES ARE INVOLVED, *Biometrika*, 34, 28-35, 10.1093/biomet/34.1-2.28, 1947.
669 Wells, B., Dolwick, P., Eder, B., Evangelista, M., Foley, K., Mannshardt, E., Misenis, C., and Weishampel,
670 A.: Improved estimation of trends in U.S. ozone concentrations adjusted for interannual variability in
671 meteorological conditions, *Atmospheric Environment*, 248, 118234,
672 <https://doi.org/10.1016/j.atmosenv.2021.118234>, 2021.
673 Zhang, G., Sun, Y., Xu, W., Wu, L., Duan, Y., Liang, L., and Li, Y.: Identifying the O₃ chemical regime
674 inferred from the weekly pattern of atmospheric O₃, CO, NO_x, and PM₁₀: Five-year observations at a
675 center urban site in Shanghai, China, *Science of The Total Environment*, 888, 164079,
676 <https://doi.org/10.1016/j.scitotenv.2023.164079>, 2023.
677 Zhu, Q., Place, B., Pfannerstill, E. Y., Tong, S., Zhang, H., Wang, J., Nussbaumer, C. M., Wooldridge, P.,
678 Schulze, B. C., Arata, C., Bucholtz, A., Seinfeld, J. H., Goldstein, A. H., and Cohen, R. C.: Direct
679 observations of NO_x emissions over the San Joaquin Valley using airborne flux measurements during
680 RECAP-CA 2021 field campaign, *Atmos. Chem. Phys. Discuss.*, 2023, 1-21, 10.5194/acp-2023-3, 2023.

681
682
683
684
685
686
687
688
689
690
691
692
693
694
695