



1 **Assessment of dicarbonyl contributions to secondary organic aerosols over China**  
2 **using RAMS-CMAQ**

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11  
12 **Abstract**

13 The concentration of secondary organic aerosol (SOA) is underestimated by one to two orders of  
14 magnitude in current model studies. Recent research suggests that the aqueous irreversible uptake  
15 of dicarbonyls contributes to the production of SOA, although few models have included this  
16 pathway. Glyoxal, an important representative component of dicarbonyls in models, is significantly  
17 underestimated. We therefore incorporated the aqueous irreversible uptake of dicarbonyls into the  
18 regional air quality modeling system RAMS-CMAQ (the Regional Atmospheric Modeling System-  
19 Community Multiscale Air Quality) to evaluate the contribution of dicarbonyls to SOA, and we then  
20 assess the impact of the underestimation of glyoxal on the production of SOA in China during two  
21 time periods: June 3 to July 11, 2014 (episode 1) and October 14 to November 14, 2014 (episode 2).  
22 When the aqueous irreversible uptake process was added, the modeled mean concentration of SOA  
23 in episode 1 increased by 1.57  $\mu\text{g}/\text{m}^3$ , which explains 15.0% of the unaccounted source of SOA. By  
24 contrast, the increase in the concentration of SOA in episode 2 was small as a result of the lower  
25 liquid water content and the lower amount of dicarbonyls produced from biogenic precursors in the  
26 fall. On this basis, when the glyoxal simulation was improved, the modeled mean dicarbonyl-  
27 derived SOA (AORGJ) increased by a factor of 2 in both episodes. AORGJ in episode 1  
28 contributed, on average, 39.0% of the total concentration of SOA and the increase in this  
29 contribution represented 26.7% of the unaccounted concentration of SOA, whereas the mean  
30 AORGJ in episode 2 accounted for 18.5% of total concentration of SOA. Based on the results for  
31 episode 1, the mean AORGJ in summer over China was generally higher in the east than in the



32 west, with the highest value (5–10  $\mu\text{g}/\text{m}^3$ ) in the areas between the lower reaches of the Yellow and  
33 Yangtze rivers and in the Sichuan Basin. The contribution of AORG CJ to the concentration of SOA  
34 varied from 10 to 90% throughout China, with the highest contributions (70–90%) in the coastal  
35 regions and offshore along the East China Sea to the South China Sea and in the southwestern  
36 regions.

37 **Keywords:** secondary organic aerosol, aqueous irreversible uptake, glyoxal, China, RAMS-CAMQ  
38

### 39 1. Introduction

40 The fine particle fraction of aerosols ( $\text{PM}_{2.5}$ , i.e., particulate matter with an aerodynamic diameter  
41  $\leq 2.5 \mu\text{m}$ ) not only absorbs and scatters sunlight in the atmosphere, resulting in a reduction in  
42 visibility and effects on the global climate (IPCC, 2007), but also has adverse effects on human  
43 health (Harrison and Yin, 2000; Poschl, 2005). Organic aerosol (OA) is a major component of fine  
44 particulate matter globally (Murphy et al., 2006; Zhang et al., 2007), typically making up 20–90%  
45 of the fine particle fraction (Kanakidou et al., 2005; Roberts et al., 2001; Zhang et al., 2007),  
46 suggesting that it has a significant effect on the characteristics and properties of fine particulate  
47 matter. Organic aerosols are formed from both primary (direct emissions) and secondary sources.  
48 Secondary organic aerosols (SOAs) are formed from precursors of volatile organic compounds  
49 (VOCs), which can be either natural or anthropogenic. However, the underestimation of organic  
50 aerosol has become a major issue in almost all current atmospheric models due to the incomplete  
51 representation of SOA (Goldstein and Galbally, 2007; Heald et al., 2005; Morris et al., 2006; Morris  
52 et al., 2005; Yu et al., 2008).

53 It has been reported that the concentration of SOA in current models is underestimated by one to  
54 two orders of magnitude (de Gouw et al., 2005; Volkamer et al., 2006). These results have motivated  
55 researchers to investigate why these models are predicting SOA concentrations so poorly.  
56 Traditionally, improvements in models have mainly concentrated on the gas-phase and derived  
57 heterogeneous formation processes, such as the formation of SOA from aromatic compounds under  
58 low- and high- $\text{NO}_x$  conditions (Ng et al., 2007), the production of SOA from the oxidation of  
59 isoprene and its subsequent growth through acid catalysis (Kroll et al., 2005, 2006; Surratt et al.,  
60 2007), the use of volatile bias set method to depict the formation of SOA instead of the two-product  
61 model (Han et al., 2016; Lane et al., 2008; Lin et al., 2016; Murphy and Pandis, 2009), and the



62 formation of SOA through the uptake of isoprene epoxides on the surface of aerosols (Hu et al.,  
63 2017;Ying et al., 2015). In spite of these researches, biases still exist between the simulations and  
64 observations.

65 Many studies have suggested that glyoxal and methylglyoxal can form SOAs through chemical  
66 reactions in cloud or fog water (e.g. Carlton et al., 2006;Ervens et al., 2004;Lim and Ziemann,  
67 2005;Loeffler et al., 2006;Warneck, 2003), or by irreversible uptake on the surface of cloud droplets  
68 and aqueous aerosols (e.g. Corrigan et al., 2008;Galloway et al., 2009;Liggio et al., 2005), which is  
69 probably a significant source of SOA (Curry et al., 2018;Ervens et al., 2014). A few studies  
70 (e.g. Carlton et al., 2010;Carlton et al., 2008;Fu et al., 2009;Li et al., 2013) have incorporated  
71 aqueous formation pathways into atmospheric models. These studies have shown that chemical  
72 reactions in cloud or fog water make negligible contribution to near-surface SOA relative to  
73 irreversible uptake on the surface of cloud droplets and aqueous aerosols, and that the aqueous  
74 formation of SOA cannot completely explain the gaps between the observations and simulations.  
75 There are still considerable uncertainties in our knowledge of the formation of SOA.

76 A series of studies (Fu et al., 2008;Liu et al., 2012;Myriokefalitakis et al., 2008;Li et al., 2018)  
77 has shown that there is a substantial underestimation in the modeled vertical column densities  
78 (VCDs) of glyoxal, but few studies have considered the impact of this underestimation on  
79 simulations of SOA concentrations. Xu et al. (2017) and Ervens et al. (2011) showed that the  
80 aqueous formation of SOA depends on the liquid water content (LWC), which varies between  
81 seasons. Previous studies, such as those of Fu et al. (2009) and Li et al. (2013), have only considered  
82 the contribution from SOA derived from the aqueous irreversible uptake of dicarbonyls (pathway  
83 M) in the summer period or over evergreen areas. In this study, we not only incorporated pathway  
84 M into the RAMS-CMAQ (the Regional Atmospheric Modeling System-Community Multiscale  
85 Air Quality) modeling system to evaluate the corresponding contribution of dicarbonyls to SOA,  
86 but also improved the simulation of glyoxal concentrations by investigating the reasons for its  
87 underestimation and assessing its impacts on the concentration of SOA during two episodes in the  
88 summer and fall.

89

## 90 **2. Model and data**

### 91 **2.1 Base model description**



92 Li et al. (2017a) used the RAMS-CMAQ modeling system to investigate the effects of  
93 underestimated aromatic VOC emissions and yields of SOAs from different gas precursors on the  
94 concentration of SOAs. Similarly, we used Version 4.7.1 of the CMAQ, which is coupled with the  
95 gas-phase photochemical mechanism SAPRC99 (1999 Statewide Air Pollutant Research Center  
96 (Carter, 2000) and Version 5 of the aerosol module (AERO5) (Byun and Schere, 2006;Foley et al.,  
97 2010). There are three major formation pathways for SOAs in this version, which is based on the  
98 two-product approach. The first pathway is the equilibrium partition of semi-volatile products  
99 formed from the oxidation of seven VOC precursors: long-chain alkanes, benzene (BNZ), high-  
100 yield aromatics (mainly toluene, ARO1), low-yield aromatics (mainly xylene, ARO2), isoprene  
101 (ISOP), monoterpene (TRP) and sesquiterpenes (SESQ). The second pathway is the oligomerization  
102 of semi-volatile SOAs formed through the first pathway (Kalberer et al., 2004), namely aging  
103 process. The third pathway is the formation of SOA via the in-cloud oxidation of glyoxal (GLY) and  
104 methylglyoxal (MGLY) (Carlton et al., 2008), both of which represent dicarbonyls in the model.  
105 The details of these formation pathways are given in Carlton et al. (2010).

106 The meteorological fields used to drive CMAQ are obtained from RAMS, which has been  
107 described in detail by Cotton et al. (2003). The National Centers for Environmental Prediction  
108 reanalysis datasets are served as the initial meteorological fields input into RAMS. The boundary  
109 conditions used for the RAMS computations include the weekly average sea surface temperature  
110 and the monthly measured snow cover. The final modeled results are output through the four-  
111 dimensional data assimilation mode using nudging analysis.

112 The emission sources are derived from several different inventories. Anthropogenic emissions  
113 (Li et al., 2017b)—including SO<sub>2</sub>, NO<sub>x</sub>, CO, black carbon, non-methane VOCs, organic carbon,  
114 NH<sub>3</sub> and other particulate matter—are obtained from the monthly emissions inventory of 2012.  
115 There are five emission sectors (the power, industrial, residential, transportation and agricultural  
116 sectors) in the inventory with a spatial resolution of 0.25° × 0.25° (see [www.meicmodel.org](http://www.meicmodel.org)). The  
117 Model of Emissions of Gases and Aerosols from Nature (MEGAN) (Guenther et al., 2012) provides  
118 the biogenic emissions. The emissions from open biomass burning are derived from the Global Fire  
119 Emissions Database, Version 4 (Randerson et al., 2015), whereas the information of monthly  
120 lightning NO<sub>x</sub> is obtained from the Global Emissions Inventory Activity with a spatial resolution of  
121 1° × 1° (Benkovitz et al., 1996). The emissions of NO<sub>x</sub> from the soil are derived from the Regional



122 Emission inventory in ASia, Version 2.1, with a spatial resolution of  $0.25^\circ \times 0.25^\circ$  (Kurokawa et al.,  
123 2013). The online dust and sea salt emissions are calculated using the empirical model developed  
124 by Han et al. (2004) and the scheme of Gong (2003), respectively. The calculation of the model  
125 boundary conditions is supported by the Model of Ozone and Related Tracers, Version 4 (MOZART-  
126 4) (Emmons et al., 2010).

127 The model domain is divided into  $105 \times 86$  grid cells with the center located at ( $35^\circ$  N,  $110^\circ$  E)  
128 (Fig. 1). The map projection is rotated polar-stereographic with a horizontal resolution of 64 km.  
129 The vertical simulation region is unequally spaced from the ground surface to  $\sim 23$  km. There are 25  
130 layers in the  $\sigma_z$  coordinate system in RAMS, with nine layers located in the lowest 2 km to resolve  
131 the planetary boundary layer. The vertical dimension of CMAQ is divided into 15 layers, with the  
132 lowest seven layers the same as those in RAMS. We focused on the area outlined by the blue box in  
133 Fig. 1.

134

## 135 2.2 Adding the aqueous irreversible uptake process

136 Pathway M was implemented in the cloud module of CMAQ following the standard reaction  
137 probability formulation of Jacob (2000). In this formulation, the first-order rate constant  $k$  for the  
138 chemical loss of a gas-phase species to the aerosols or cloud droplets through molecular diffusion  
139 and free collision is given by

$$140 \quad k = \left( \frac{a}{D_g} + \frac{4}{v\gamma} \right)^{-1} A \quad (1)$$

141 where  $D_g$  is the gas-phase molecular diffusion coefficient,  $a$  is the radius of the aerosol particle or  
142 cloud droplet,  $v$  is the mean molecular speed,  $\gamma$  is the reaction uptake coefficient when a collision  
143 occurs between a gas-phase molecule and the aqueous surface, and  $A$  is the aqueous particle surface  
144 area per unit volume of air.

145 In a similar manner to Fu et al. (2008), diffusion limitation ( $D_g A/a$ ) is applied in the model to  
146 account for the uptake of gas by cloud droplets. As we are mainly concerned with continental regions,  
147 the cloud droplet surface area is calculated from the LWC in the cloudy fraction by assuming an  
148 effective droplet radius of  $6 \mu\text{m}$  (Li et al., 2013).  $D_g$  is chosen as the typical tropospheric value given  
149 by Jacob (2000). The collision limitation ( $v\gamma A/4$ ) is then used to compute the uptake of gas by  
150 aqueous aerosols as reported by Ying et al. (2015). The value of  $\gamma = 2.9 \times 10^{-3}$  adopted here for both



151 glyoxal and methylglyoxal is from Liggio et al. (2005).

152

### 153 **2.3 Observational data**

154 To assess the contribution of dicarbonyl species to the concentration of SOA, we used the observed  
155 concentration of SOA in Beijing during the summer and fall of 2014 measured by Xu et al. (2017)  
156 with an Aerodyne high-resolution time-of-flight aerosol mass spectrometer. Samples were taken  
157 from June 3 to July 11, 2014 and October 14 to November 14, 2014 at the Institute of Atmospheric  
158 Physics. More detailed information about the data has been reported by Xu et al. (2017). The  
159 simulation periods are from May 22 to July 11 and from October 1 to November 14, with the first  
160 12 days as the spin-up time. To evaluate the reasonability in simulating the formation processes of  
161 SOAs in the aqueous phase, the corresponding cloud water path (CWP) data measured by the  
162 MODerate Resolution Imaging Spectroradiometer (MODIS) was also obtained from the website  
163 <http://ladsweb.modaps.eosdis.nasa.gov/api/v1/productPage/product=MYDATML2>.

164 Liu et al. (2012) and Li et al. (2018) have suggested that the underestimation of glyoxal  
165 concentrations in simulations is related to the underpredicted emission of aromatic compounds and  
166 the molar yields of glyoxal from the precursors. Therefore the biases in the emission of aromatic  
167 compounds needs to be evaluated through a comparison of the observed and simulated  
168 concentrations of aromatic compounds. The observed data were collected at 14:00 local standard  
169 time every Thursday by gas chromatography–mass spectrometer at Beijing, Xinglong and Yucheng  
170 (Fig. 1). Sun et al. (2016) and Wu et al. (2016) have presented the detailed information.

171 To evaluate the performance of our model, we also compared the simulated PM<sub>2.5</sub> concentrations  
172 and several meteorological factors with the results measured over 12 stations (Fig. 1) during the two  
173 episodes. The observed meteorological data were derived from the Meteorological Information  
174 Comprehensive Analysis and Process System (MICAPS) dataset, whereas the Chinese National  
175 Environmental Monitoring Center provided the measured concentrations of PM<sub>2.5</sub>.

176

## 177 **3. Results and discussions**

### 178 **3.1 Model evaluation**

179 Table 1 shows statistical results of the meteorological parameters (i.e. temperature, relative humidity,  
180 wind speed and wind direction) and PM<sub>2.5</sub> concentrations in the two analyzed episodes, where:  $N$  is



181 the total number of samples; IOA is the index of agreement, which can synthetically reflect the  
182 combination of the modeled value and variable tendency being good or bad (Willmott, 1981);  $C_{\text{mod}}$   
183 and  $C_{\text{obs}}$  are the average values of modeled and observed results, respectively; MB and FB are the  
184 mean and fractional bias, respectively; RMSE and FE are the root-mean-square and fractional error,  
185 respectively; and  $R$  is the correlation coefficient between the observed and simulated results. The  
186 calculations of these statistical parameters can be found in Juda-Rezler et al. (2012).  $P_{22.5^\circ}$  and  $P_{45^\circ}$   
187 represent the proportions of compared results that the absolute biases between the simulated and  
188 measured wind directions are within  $22.5^\circ$  and  $45^\circ$ , respectively (Li et al., 2017a).

189 There are inevitably some biases in the simulated meteorological parameters relative to the  
190 observations due to the limited model resolution and system errors. Nevertheless, the model  
191 reproduces the magnitude and variation trend of the temperature and relative humidity fairly well  
192 (Figs. S1 and S2), with IOAs for temperature of 0.91 and 0.95 in the two episodes and IOAs of 0.91  
193 and 0.88 for the relative humidity, comparable with the results of Li et al. (2012) and Wang et al.  
194 (2014). The mean biases of temperature ( $-0.83^\circ\text{C}$ ,  $-0.68^\circ\text{C}$ ) and relative humidity (1.53%,  $-0.05\%$ )  
195 are small in the two episodes. The RMSEs of temperature and relative humidity are comparable  
196 with the results of Gao et al. (2016) and Zhang et al. (2012), implying a reasonable performance of  
197 the simulation. Although the capture of the variable tendency of wind speed (Fig. S3) by the model  
198 is rather poor, the small values for the mean bias ( $-0.40$  and  $-0.22$  m/s) and the high IOAs (0.61 and  
199 0.69) indicate a better performance than that reported by Feng et al. (2016) and Wang et al. (2014)  
200 in simulating the wind speed. The RMSE of wind speed shows that the simulated results achieve  
201 the criteria for a good performance ( $|\text{RMSE}| \leq 2$  m/s) given by Emery et al. (2001). For the wind  
202 direction (Fig. S4),  $P_{22.5^\circ}$  and  $P_{45^\circ}$  are 32.12 and 54.52% in episode 1 and 34.01 and 57.41% in  
203 episode 2, indicating the reasonable simulation of wind direction. Therefore the model provides a  
204 reasonable meteorological field for the subsequent simulations.

205 The modeled  $\text{PM}_{2.5}$  concentrations are generally higher than the observed values in the two  
206 episodes over the 12 stations (Fig. S5). Except for the limited resolution of the model, the  
207 uncertainties of the emission inventory and the effects of the background transport may also  
208 contribute to the overestimation. However, the correlation coefficients of  $\text{PM}_{2.5}$  are 0.50 and 0.56 in  
209 the two episodes, respectively, and the IOAs are 0.64 and 0.70 in the two episodes, respectively,  
210 indicating good capture of the variable tendency and magnitude by the model. Both the fractional



211 error and the fractional bias for PM<sub>2.5</sub> in the two episodes fulfill the performance criteria ( $FE \leq$   
212  $75\%$ ,  $|FB| \leq 60\%$ ) given by Boylan and Russell (2006), implying a good performance in  
213 simulating PM<sub>2.5</sub>. These results show that the model gives a reasonable simulation of the chemical  
214 species.

215 As described in Section 2.2, pathway M is related to the cloud water content. Thus the modeled  
216 and satellite-observed mean CWP were compared during the two analyzed episodes to evaluate the  
217 reasonability of the simulation results. Fig. 2a and 2b show that in episode 1, during one summer  
218 period, the highest observed CWP (400–500 g/m<sup>2</sup>) mainly appears over the Yellow Sea and Bohai  
219 Sea and in parts of southeastern and southern China, whereas the observed second highest values  
220 (200–300 g/m<sup>2</sup>) are concentrated in from the Qinghai–Tibetan Plateau to the North China Plain  
221 along the Yellow River Basin. The modeled high CWP (100–400 g/m<sup>2</sup>) appears in the same regions  
222 as in the observational dataset. In episode 2, during one fall period (Fig. 2c and 2d), the observed  
223 highest (300–500 g/m<sup>2</sup>) and second highest (200–300 g/m<sup>2</sup>) CWP values are mainly concentrated in  
224 the Hetao area and the regions including the northeastern China and Inner Mongolia, respectively.  
225 The modeled values (50–150 g/m<sup>2</sup>) are also high over these regions.

226 Overall, there are obvious biases in the numerical values between the observed and simulated  
227 CWP. This is a result of the large uncertainties in both the model simulations and the satellite  
228 retrievals of clouds, which lead to uncertainties in the simulations of the concentrations of SOAs.  
229 However, the mean distribution patterns of the simulated CWP during the two episodes are similar  
230 to the observational results, indicating few impacts on the simulated distribution of SOA. Both the  
231 simulated and observed CWP in episode 1 are higher than in episode 2, also implying differences in  
232 the simulation of SOA between the two episodes.

233

### 234 **3.2 Model results and analyses**

235 As described in Section 2.3 and with reference to previous research, the underestimation of glyoxal  
236 concentrations may result from the underprediction of the emissions of aromatic compounds and  
237 the molar yields of glyoxal from the precursor molecules. It is therefore necessary to evaluate the  
238 emissions of aromatic compounds during the analyzed episodes before designing and implementing  
239 the model case studies. For this purpose, comparisons were made between the simulated and  
240 observed concentrations of aromatic compounds in a similar manner to the work of Zhang and Ying



241 (2011).

242 Fig. 3 presents the ratio of the observed to predicted (O/P ratio) concentrations of aromatic  
243 compounds for the original emissions (Fig. 3a and 3c) and a three-fold increase in the emissions of  
244 aromatic compounds (Fig. 3b and 3d) during the two episodes. Fig. 3a shows that there are biases  
245 between the observed and simulated concentrations of aromatic compounds in episode 1 at the  
246 original emission rates of the aromatic compounds. The O/P ratios are more concentrated for ARO1  
247 than for ARO2, with both the mean and median O/P values being close to 2. The range of O/P values  
248 is large for ARO2 and there is an order of magnitude difference between the lowest and highest  
249 ratios. Both the mean and median O/P ratios of ARO2 are about 20. These results show that there  
250 are underestimations in the emissions of ARO1 and ARO2. In view of the limitations of the amount  
251 of observed data and the nonlinear relationship between emissions and the concentrations (Li et al.,  
252 2018), the underestimation in the emission rates has been conservatively estimated to be a factor of  
253 three. As a result, when the emission rates of aromatic compounds are increased (Fig. 3b), the ratios  
254 of ARO1 and ARO2 in episode 1 are more concentrated, especially ARO2. The mean and median  
255 ratios of ARO1 are both around 1, whereas the two ratios for ARO2 decrease from 20 to 6. The  
256 results for episode 2 are similar to those in episode 1. Fig. 3c shows that the original O/P ratio of  
257 ARO1 varies between 1 and 3.5, with a mean value close to 2, whereas the ratio of ARO2 ranges  
258 from 1 to 5, with a mean ratio of about 4, both of which indicate unaccounted emissions of aromatic  
259 compounds. Fig. 3d shows a clear decrease in the biases of the observed and simulated  
260 concentrations of aromatic compounds. The O/P ratios of ARO1 and ARO2 are concentrated and  
261 vary between 1 and 2, with a mean ratio close to 1. However, it is difficult to determine whether the  
262 factor of 3 is the actual underestimation in the emission of aromatic compounds as a result of its  
263 dependence on space and time. For convenience, a factor of 3 is chosen here as a uniform scale with  
264 which to assess the unaccounted emissions of aromatic compounds.

265 Three sensitivity simulation case studies were designed based on these results. Case 0 is the base  
266 example and was performed with the standard model. Case 1 was run with the modified model  
267 incorporating pathway M. By combining our results with those of Li et al. (2018), case 2 was  
268 designed to take the effects of the underestimations of glyoxal concentrations into considerations  
269 (i.e. a three-fold increase in the emissions of aromatic compounds and a five-fold increase in the  
270 molar yield of glyoxal from isoprene) based on the results of case 1.



271 Fig. 4 compares the hourly concentrations of the simulated and observed SOAs during the two  
272 analyzed episodes; the corresponding statistical parameters are listed in Table 2. The concentrations  
273 of SOAs are measured from PM<sub>1</sub>. The observed PM<sub>1</sub>/PM<sub>2.5</sub> ratio of 0.77 (Xu et al., 2015) was used  
274 to convert the observed concentrations for comparison with the simulated results for PM<sub>2.5</sub>.

275 In case 0, the SOA concentrations in episode 1 (Fig. 4a) are significantly underestimated by an  
276 average factor of 5.7, with the differences being as high as a multiple of  $\geq 60$ . As a result of the  
277 impacts of uncertainties in the meteorological fields and emissions, the variation trend of the  
278 concentration of SOA was not well captured by the model ( $R = 0.21$ ). Similarly, although the  
279 variation of SOA was reproduced well in episode 2 ( $R = 0.83$ ) (Fig. 4b), the concentration of SOA  
280 was still underpredicted by an order of magnitude. The biases between the observed and simulated  
281 concentrations of SOAs decreased in case 1 when pathway M was added to the model, especially in  
282 episode 1. Fig. 4c shows that the mean concentration of SOA in case 1 increased by 1.57  $\mu\text{g}/\text{m}^3$   
283 during episode 1, explaining 15.0% of the unaccounted sources of SOAs. The decreased mean bias  
284 and RMSE and the slightly increased IOA also indicate the more reasonable descriptions of the  
285 formations of SOAs. In episode 2, the SOA formed through pathway M contributed less to the total  
286 concentration of SOA (Fig. 4d). The mean SOA concentration during episode 2 reached 1.31  $\mu\text{g}/\text{m}^3$ ,  
287 an increase of only 0.26  $\mu\text{g}/\text{m}^3$ . The other statistical parameters showed little change, indicating the  
288 small contribution of dicarbonyls through pathway M to the concentration of SOA in this episode.  
289 The larger contribution of dicarbonyls from pathway M to the concentration of SOA in episode 1 is  
290 attributed to the higher LWC and the larger amount of dicarbonyls produced from biogenic  
291 precursors in the summer than in the fall. When the impact of the underestimation of glyoxal on the  
292 concentration of SOA is taken into consideration, the concentration of SOA clearly increases in case  
293 2. Fig. 4e and 4f show that the mean concentration of SOA in episode 1 is significantly improved  
294 by a factor of 2.3, whereas the increase in episode 2 is a factor of 1.8. The statistical parameters (e.g.  
295 IOA,  $R$  and RMSE) also show the better performance of case 2, indicating a more realistic and  
296 reasonable representation of the formation of SOA. Aromatic compounds are not only the precursors  
297 of glyoxal, but are also a precursor of SOA in the gas phase. A compositional analysis is therefore  
298 required to evaluate the individual contributions from dicarbonyls in case 2.

299 Fig. 5 shows the mean contributions from different sources of SOAs in the three sensitivity case  
300 studies during the two analyzed episodes. AORGJ, dicarbonyl-derived SOAs, contributes little in



301 case 0 during the two episodes. The mean contribution of AORG CJ to the total concentration of  
302 SOAs in episode 1 is 2.39%, larger than in episode 2 (0.89%). As a result of the greater amount of  
303 emissions from biogenic sources in summer, SOAs formed from biogenic precursors  
304 (AISOP+ATRP+ASQT) contribute more (26.3%) in episode 1 than that in episode 2 (13.5%),  
305 whereas the contributions from anthropogenic precursors (AALK+ATOL+AXYL+ABNZ) are  
306 comparable between the two episodes. When pathway M is included in the model, the contribution  
307 from AORG CJ to SOA in case 1 clearly increases and reaches 37.1% in episode 1 and 19.3% in  
308 episode 2. The contribution from AORG CJ in episode 1 is higher than the contributions from  
309 anthropogenic (26.9%) and biogenic (16.6%) gas precursors, indicating the significant contributions  
310 from dicarbonyls through pathway M to the formations of SOAs in summer. When the impact of  
311 the underestimation of glyoxal is taken into consideration in case 2, the mean concentration of  
312 AORG CJ in both episodes increases by nearly a factor of 2 relative to case 1. In episode 1, AORG CJ  
313 ( $2.85 \mu\text{g}/\text{m}^3$ ) accounts for 39.0% of the total SOA, exceeding the sum of the contributions from  
314 ARO1 (ATOL) and ARO2 (AXYL) (26.0%), and indicating the dominant contribution of aqueous-  
315 phase process to the concentration of SOA in summer. The increase in AORG CJ compensates for  
316 about 26.7% of the unaccounted sources of SOAs. In episode 2, AORG CJ ( $0.56 \mu\text{g}/\text{m}^3$ ) contributes  
317 18.5% to the total concentration of SOA, which is much less than the sum of ATOL and AXYL  
318 (38.3%), implying the dominant contributions of aromatic compounds to the concentrations of  
319 SOAs. The different contributions of AORG CJ in case 2 during the two episodes can be attributed  
320 to the different LWC and different amount of dicarbonyls produced from biogenic precursors.

321 It is clear that the biases between the observed and simulated concentrations of SOAs decrease  
322 when the contributions of unaccounted dicarbonyls to the concentrations of SOAs are considered,  
323 especially in summer. However, the sources of unaccounted SOAs cannot be explained completely.  
324 As a result of uncertainties in the description of known SOA formation processes and missing  
325 pathways that have not been included in the model—for example, there are many uncertainties in  
326 glyoxal simulations (Li et al. (2018)). There are also many uncertainties in incorporating pathway  
327 M into the model, such as the effective radius of cloud droplets (the model only considers the  
328 empirical value over continental regions), the diffusion coefficient (which does not consider the  
329 impacts of temperature, molecular weight and air density), the reaction uptake coefficient (updated  
330 values over disparate land surfaces are discussed in Curry et al. (2018)) and the liquid water content



331 in clouds (the most uncertain parameter in Fig. 2). This study considers the formation of SOA in  
332 clouds through pathway M, but neglects processes occurring on the surface of fog droplets or  
333 aqueous aerosols near the ground, which may be important sources of SOAs in fall and winter. Other  
334 pathways for the formations of SOAs, such as the uptake of isoprene epoxides on the surface of  
335 aerosols (Lal et al., 2012; Lin et al., 2013), the aging mechanism of semi-volatile primary organic  
336 aerosols (Shrivastava et al., 2008) and the oxidation of primary semi- and intermediate VOCs, have  
337 not been considered in this model.

338 To distinguish the contribution of dicarbonyls to the concentration of SOA over China, the  
339 distributions of dicarbonyl-derived SOAs and their contributions to SOAs (AORG CJ/SOA) over  
340 China in cases 0 and 2 were analyzed. The contributions from AORG CJ were not clear in episode  
341 2, thus only the results for episode 1 are here.

342 Fig. 6a and 6b show the mean concentration of AORG CJ in cases 0 and 2, respectively, during  
343 episode 1. For the base case, the concentration of AORG CJ over China is  $\leq 0.2 \mu\text{g}/\text{m}^3$ . The  
344 concentration of AORG CJ is higher ( $0.1\text{--}0.2 \mu\text{g}/\text{m}^3$ ) in the areas between the lower reaches of the  
345 Yellow and Yangtze rivers and in the Sichuan Basin, probably due to the higher LWC and a greater  
346 number of sources of dicarbonyls over these areas, as discussed in Section 3.1 and by Li et al. (2018).  
347 The concentration of AORG CJ is clearly improved in case 2, when pathway M is added and the  
348 impact of the underestimation of glyoxal is considered. Overall, the concentration of AORG CJ is  
349 higher in eastern China than in the west. The concentrations of AORG CJ vary from 2 to  $5 \mu\text{g}/\text{m}^3$   
350 over central and eastern China, with the highest value ( $5\text{--}10 \mu\text{g}/\text{m}^3$ ) concentrated in the same areas  
351 as in case 0. The concentration of AORG CJ in western China is  $\leq 1 \mu\text{g}/\text{m}^3$ , with the lowest value ( $\leq$   
352  $0.1 \mu\text{g}/\text{m}^3$ ) in the northeast, probably because there are few sources of dicarbonyls here.

353 Fig. 6c and 6d show the spatial distribution of the mean AORG CJ/SOA in cases 0 and 2 during  
354 episode 1, respectively. Fig. 6c shows that the AORG CJ fraction over China is  $\leq 10\%$ , except in  
355 Yunnan Province and some parts of the South China Sea, where AORG CJ/SOA reaches 10–20%.  
356 When the contributions of dicarbonyls from pathway M and improved glyoxal to SOA are  
357 considered, there is a large increase in AORG CJ/SOA over these regions. Fig. 6d shows that the  
358 contribution of AORG CJ to SOA varies from 10 to 90% throughout China. In central and eastern  
359 areas, the fraction of AORG CJ is generally in the range of 40–60% and up to 70–90% in the coastal  
360 regions and offshore from the East China Sea to the South China Sea. The fraction of AORG CJ in



361 the west is relatively low and usually  $\leq 40\%$ . However, the contribution of AORG CJ to SOA is up  
362 to 80% in southwestern regions (e.g. Yunnan Province and parts of Sichuan Province).

#### 363 4. Conclusions

364 The RAMS-CMAQ modeling system was used to assess the contributions of dicarbonyls from the  
365 aqueous irreversible uptake process and unaccounted sources of glyoxal to the concentrations of  
366 SOAs during two episodes from June 3 to July 11, 2014 and October 14 to November 14, 2014.  
367 Comparisons between the observed and simulated concentrations of SOAs from three sensitivity  
368 groups showed different improvements in the SOA simulations with the inclusion of pathway M  
369 and consideration of the underestimation of glyoxal in the two episodes. Due to the high LWC and  
370 large amount of dicarbonyls produced from biogenic precursors in summer, the contributions of  
371 dicarbonyls were greater in episode 1 than in episode 2. When pathway M was added in case 1, the  
372 modeled mean concentration of SOA in episode 1 increased by  $1.57 \mu\text{g}/\text{m}^3$ , explaining about 15.0%  
373 of the unaccounted SOA sources, whereas there was only a  $0.26 \mu\text{g}/\text{m}^3$  increase in the mean result  
374 during episode 2. When the impacts of glyoxal underestimation were taken into consideration in  
375 case 2, the modeled mean AORG CJ in episode 1 was improved by a factor of 2 and contributed  
376 39.0% of the total concentration of SOA. The increase due to AORG CJ is equivalent to 26.7% of  
377 the unaccounted source of SOA. Whereas the mean concentration of AOGRCJ in episode 2 was  
378 only  $0.56 \mu\text{g}/\text{m}^3$  and accounted for 18.5% of total concentration of SOA. Although the contributions  
379 of dicarbonyls to SOAs are different in the two episodes, the simulated SOA results are both  
380 improved and close to the observations, indicating a more realistic description of aqueous formation  
381 of SOA.

382 The mean AORG CJ in case 2 during episode 1 was clearly improved over China relative to case  
383 0 and was generally higher in the east than in the west. The highest value ( $5\text{--}10 \mu\text{g}/\text{m}^3$ ) was seen in  
384 the areas between the lower reaches of the Yellow and Yangtze rivers and in the Sichuan Basin, as  
385 a result of the combination of a higher liquid water content and the greater number of dicarbonyl  
386 sources in these areas. As a result, the contribution of AORG CJ to the concentration of SOA in  
387 episode 1 was also improved in case 2. The fraction varied from 10 to 90% throughout China, with  
388 the highest contribution (70–90%) in the coastal regions or offshore along the East China Sea to the  
389 South China Sea in addition to the southwestern regions (e.g. Yunnan Province and parts of Sichuan  
390 Province).



391 It is clear that the contributions of dicarbonyls from pathway M and the unaccounted sources of  
392 glyoxal make a significant contribution to the concentration of SOA, especially in summer. However,  
393 there are still many uncertainties in the depictions of aqueous irreversible uptake processes and  
394 sources of glyoxal. More work is needed on the parameters used in pathway M (e.g. the effective  
395 radius of cloud droplets, the diffusion coefficient, and the reaction uptake coefficient), and to  
396 improve the accuracy of the modeled liquid water content in clouds.

397

398 *Author contribution.* In this study, J. Li designed the sensitivity experiments, developed the model code, and  
399 performed the corresponding simulations. M. Zhang co-designed the experiments and provided valuable advices  
400 about the model operations. Y. Sun carried out the measurements of SOA and provided the corresponding data for  
401 evaluating the modeled results. G. Tang and F. Wu carried out the measurements of aromatic compound  
402 concentrations and provided the corresponding data for evaluating the simulated results. Y. Xu provided valuable  
403 advice on model result analysis.

404

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641 Table 1. Statistics for the meteorological variables and PM<sub>2.5</sub> between the simulated and observed data during the  
 642 two episodes.

Variables	Episode	<i>N</i>	<i>C</i> <sub>mod</sub>	<i>C</i> <sub>obs</sub>	MB	RMSE	<i>R</i>	IOA
Temperature (°C)	1	10770	23.76	24.59	-0.83	2.58	0.86	0.91
	2	8391	12.68	13.36	-0.68	3.13	0.92	0.95
Relative humidity (%)	1	10770	69.42	67.90	1.52	13.68	0.83	0.91
	2	8389	62.81	62.86	-0.05	14.73	0.77	0.88
Wind speed (m/s)	1	10770	1.78	2.18	-0.40	1.64	0.35	0.61
	2	8391	2.05	2.27	-0.22	1.75	0.50	0.69

Variables	Episode	<i>N</i>	<i>C</i> <sub>mod</sub>	<i>C</i> <sub>obs</sub>	FB (%)	FE (%)	<i>R</i>	IOA
PM <sub>2.5</sub> (µg/m <sup>3</sup> )	1	8091	64.04	47.90	13.35	64.61	0.50	0.64
	2	7649	86.25	61.67	36.00	62.27	0.56	0.70

Variables	Episode	<i>N</i>	P <sub>22.5°</sub> (%)	P <sub>45°</sub> (%)
Wind direction (°)	1	10770	32.12	54.52
	2	8391	34.01	57.41

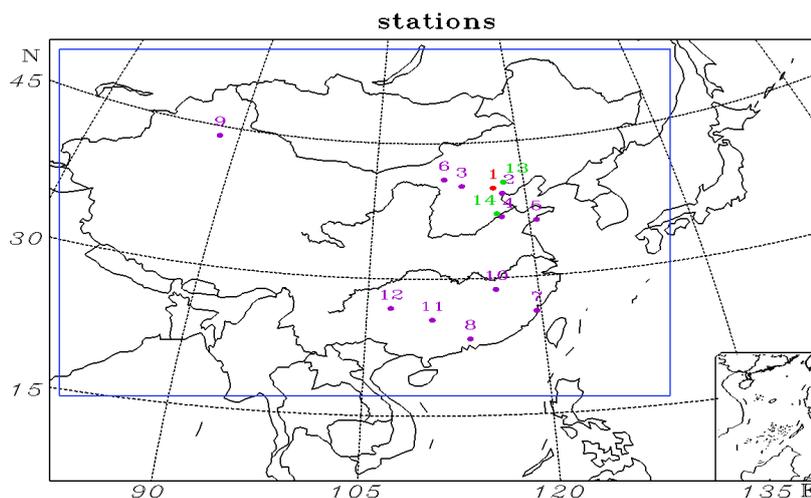
643 *N* is the total number of samples; *C*<sub>mod</sub> and *C*<sub>obs</sub> are the average value of modeled and observed results, respectively;  
 644 MB and FB are the mean and fractional bias, respectively; RMSE and FE are the root-mean-square and fractional  
 645 error, respectively; IOA is the index of agreement; *R* is the correlation coefficient between the observed and simulated  
 646 results; P<sub>22.5°</sub> and P<sub>45°</sub> represent the proportions of compared results that the absolute biases between the simulated  
 647 and measured wind directions are within 22.5° and 45°, respectively.

648

649 Table 2. Performance statistics of the modeled and observed SOA concentrations (µg/m<sup>3</sup>).

Case	Episode	<i>N</i>	<i>C</i> <sub>model</sub>	<i>C</i> <sub>obs</sub>	MB	RMSE	<i>R</i>	IOA
0	1	822	2.21	12.69	-10.48	13.48	0.21	0.46
	2	737	1.05	17.69	-16.64	23.52	0.83	0.45
1	1	822	3.78	12.69	-8.91	12.16	0.30	0.48
	2	737	1.31	17.69	-16.38	23.10	0.80	0.46
2	1	822	7.33	12.69	-5.36	10.10	0.27	0.50
	2	737	3.02	17.69	-14.67	20.79	0.80	0.51

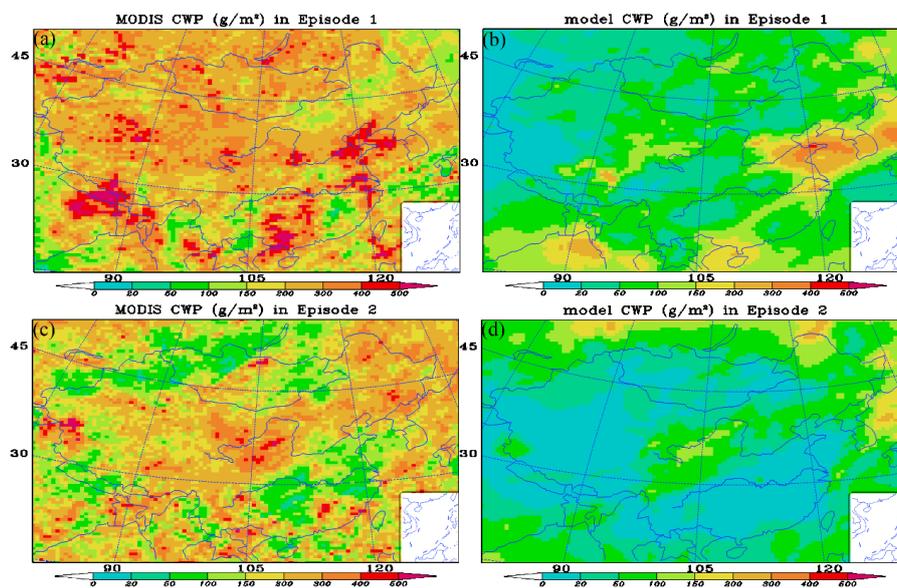
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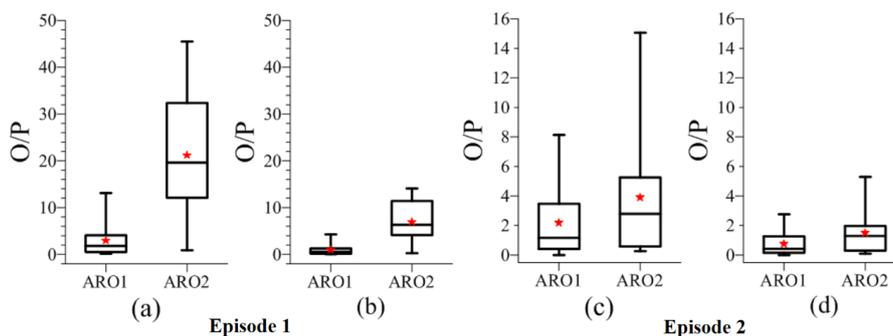
652 Fig. 1. Geographical locations of the measurement stations in the model domain. 1: Beijing; 2: Tianjin; 3: Datong;  
653 4: Jinan; 5: Qingdao; 6: Hohhot; 7: Fuzhou; 8: Guangzhou; 9: Urumqi; 10: Nanchang; 11: Guilin; 12: Guiyang; 13:  
654 Xinglong; 14: Yucheng. The different colors denote which variables are compared at each station: purple:  
655 meteorological parameters; green: concentrations of aromatics; red: both meteorological parameters and aromatic  
656 compound concentrations.

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659 Fig. 2. Observed and simulated distributions of the mean cloud water path (CWP) during the two episodes (left-hand  
660 panels: MODIS observation; right-hand panels: results of simulations).



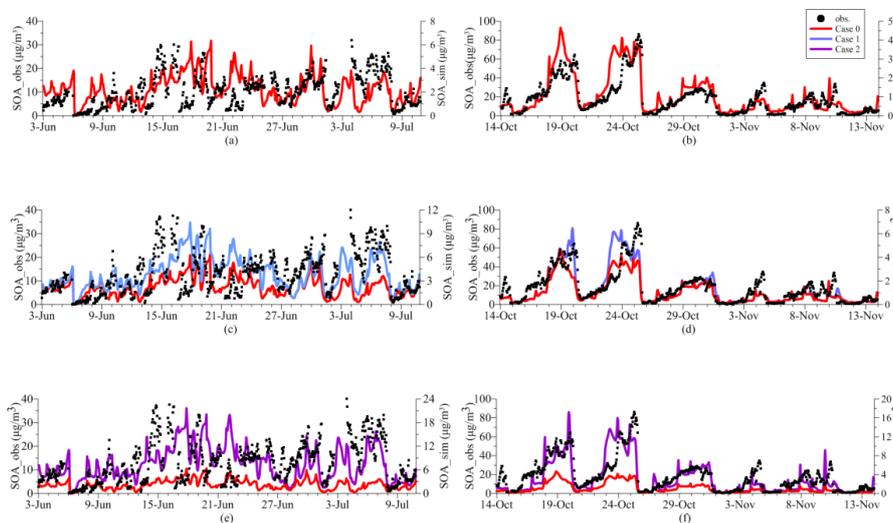
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662 Fig. 3. Box-whisker plot of the observed to predicted (O/P) ratios of aromatic compounds during the two analyzed

663 episodes at three sites (Beijing, Xinglong and Yucheng). (a, c): Original emissions of aromatic compounds; and (b,

664 d): a three-fold increase in the emissions of aromatic compounds. The red stars show the mean O/P ratios.

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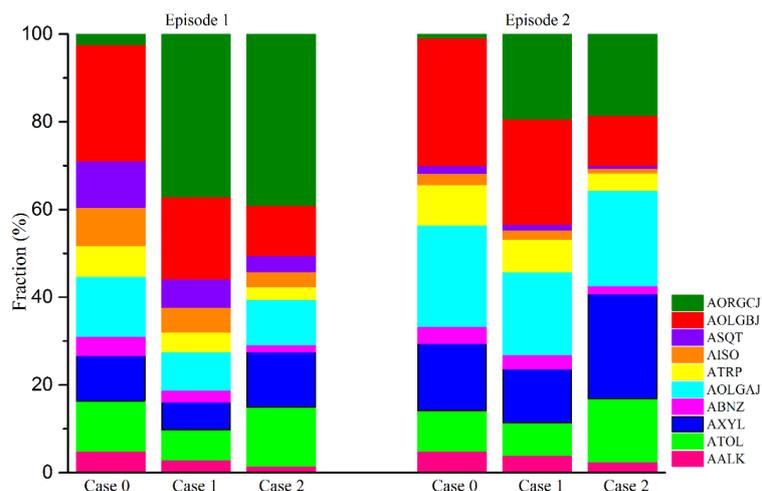


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667 Fig. 4. Hourly concentrations of the observed and simulated near-surface SOA concentrations in the two analyzed

668 episodes. (a, b) Case 0 is base run; (c, d) case 1 is run with the incorporation of the aqueous irreversible uptake of

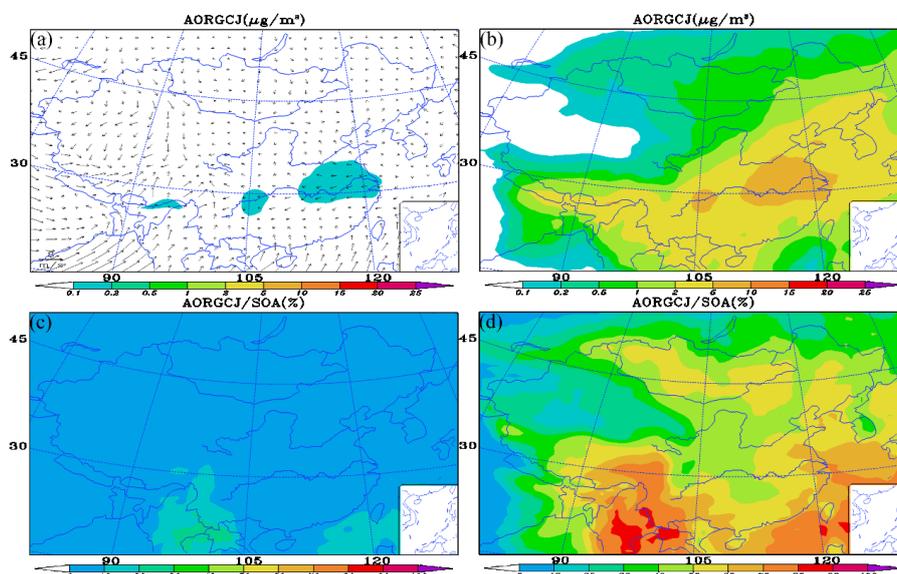
669 dicarbonyls; and (e, f) case 2 is run with the glyoxal underestimation taken into consideration based on case 1.



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671 Fig. 5. Mean contributions from different sources of SOAs in the three sensitivity case studies during the two  
 672 analyzed episodes. The compositions represent the SOA formed from long-chain alkanes (AALK), high-yield  
 673 aromatic compounds (ATOL), low-yield aromatic compounds (AXYL), benzene (ABNZ), monoterpenes (ATRP),  
 674 isoprene (AISO) and dicarbonyls (AORGCJ), as well as aged anthropogenic (AOLGAJ) and biogenic SOA  
 675 (AOLGBJ). Case 0 is the base run; case 1 is run with the incorporation of the aqueous irreversible uptake of  
 676 dicarbonyls; and case 2 is based on case 1, but taking into consideration the underestimation of glyoxal.

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678

679 Fig. 6. Modeled distributions of the mean (a) wind field with AORGCJ in case 0, (b) AORGCJ in case 2, (c)



680 AORGCI/SOA in case 0, (d) AORGCI/SOA in case 2, over the regions during episode 1. Case 0 is the base run;  
681 case 2 is run taking into consideration the incorporation of the aqueous irreversible uptake of dicarbonyls and the  
682 underestimation of glyoxal.