



Effects of Secondary Organic Aerosol Water on fine PM levels and composition over US

3 4

5 Stylianos Kakavas^{1,2}, Spyros N. Pandis^{1,2} and Athanasios Nenes^{1,3}

- ¹Institute of Chemical Engineering Sciences, Foundation for Research and
 7 Technology Hellas, Patras, Greece
- 8 ²Department of Chemical Engineering, University of Patras, Patras, Greece
- 9 ³School of Architecture, Civil and Environmental Engineering, École Polytechnique
- 10 Fédérale de Lausanne (EPFL), Switzerland11
- 12 *Correspondence to*: Spyros N. Pandis (spyros@chemeng.upatras.gr) and Athanasios13 Nenes (athanasios.nenes@epfl.ch).
- 14

Abstract. Water is a key component of atmospheric aerosol, affecting many aerosol 15 processes including gas/particle partitioning of semi-volatile compounds. Water 16 related to secondary organic aerosol (SOAW) is often neglected in atmospheric 17 18 chemical transport models and is not considered in gas-to-particle partitioning calculations for inorganic species. We use a new inorganic aerosol thermodynamics 19 20 model, ISORROPIA-lite, which considers the effects of SOAW, to perform chemical 21 transport model simulations for a year over the continental United States to quantify its effects on aerosol mass concentration and composition. SOAW can increase 22 23 average fine aerosol water levels up to a factor of two when secondary organic aerosol 24 (SOA) is a major PM₁ component. This is often the case in the south-eastern U.S where SOA concentrations are higher. Although the annual average impact of this 25 added water on total dry PM₁ concentrations due to increased partitioning of nitrate 26 and ammonium is small (up to 0.1 μ g m⁻³), total dry PM₁ increases of up to 2 μ g m⁻³ 27 28 (with nitrate levels increases up to 200%) can occur when RH levels and PM_1 concentrations are high. 29

30

31 **1. Introduction**

Fine atmospheric particulate matter with aerodynamic diameter smaller than 2.5 μm
(PM_{2.5}) has adverse effects on public health, climate and ecosystem productivity (Pye
et al., 2020; Baker et al., 2021; Guo et al., 2021). PM_{2.5} is composed of thousands of
organic compounds, black carbon (BC), and inorganic components such as sulfate





(SO₄²⁻), nitrate (NO₃⁻), ammonium (NH₄⁺) and chloride (Cl⁻) (Seinfeld and Pandis,
2006). Potassium (K⁺) levels can also be significant during biomass burning events
(Zhang et al., 2015; Pye et al., 2020). Ambient aerosol is mostly composed of water
which is determined by the chemical equilibrium of water vapor with the aerosol
constituents (Liao and Seinfeld, 2005; Carlton and Turpin, 2013; Guo et al., 2015;
Bougiatioti et al., 2016; Nguyen et al., 2016; Guo et al., 2017; Song et al., 2018; Pye
et al., 2020).

The hygroscopicity parameter (κ), which expresses the ability of a PM 43 44 component to absorb water, is an effective approach for the parameterization of the water uptake of atmospheric PM that is a mixture of organic and inorganic species 45 (Petters and Kreidenweis, 2007). Although organic aerosol (OA) is less hygroscopic 46 47 than inorganic salts, it can still contribute significantly to the total aerosol water (Guo 48 et al., 2015; Bougiatioti et al., 2016; Jathar et al., 2016) or can even become the dominant contributor at lower ambient relative humidity (Jin et al., 2020). Previous 49 studies have demonstrated that secondary organic aerosol (SOA) is a lot more 50 51 hygroscopic than primary organic aerosol (POA) and is mainly responsible for the 52 corresponding OA water (Koehler et al., 2009; Jathar et al., 2016).

53 SOAW can enhance secondary inorganic aerosol concentrations assisting in 54 their partitioning in the particulate phase to satisfy equilibrium. However, such effects 55 are not considered in thermodynamic modules used for the simulation of gas-toparticle partitioning of inorganic species in chemical transport models. Evidence 56 57 exists however that fine aerosol nitrate and ammonium concentrations can increase in areas with high organic aerosol and RH levels (Kakavas et al., 2022). The importance 58 of these SOAW impacts on secondary aerosol formation has not been systematically 59 60 studied and is the focus of this work.

We use a new aerosol thermodynamics model, ISORROPIA-lite (Kakavas et al., 2022), to simulate SOAW effects on the partitioning of the inorganic components, for a year over the continental United States. The model performance has been evaluated for fine PM and its components for the examined period by Skyllakou et al. (2021). The aim of our work is to quantify the SOAW contribution to the total fine PM water and to study its effects on inorganic aerosol thermodynamics and total dry fine PM levels and composition.

68





70 2.1 ISORROPIA-lite

71 ISORROPIA-lite is a lean and accelerated version of the widely used ISORROPIA-II (Fountoukis and Nenes, 2007) aerosol thermodynamics model. It assumes that the 72 73 aerosol exists only in the metastable state at low RH and the activity coefficients of 74 ionic pairs are always obtained from precalculated look-up tables. It estimates aerosol water associated with each one of the aerosol components. Furthermore, 75 ISORROPIA-lite has an important additional feature compared to ISORROPIA-II, as 76 77 it considers the effects of SOAW on inorganic aerosol thermodynamics. The resulting 78 increase of the total water mass drives more of the water-soluble gaseous species to 79 the particle phase to satisfy equilibrium. SOAW, WSOA, in ISORROPIA-lite is 80 calculated using the well-established κ -Kohler theory of Petters and Kreidenweis 81 (2007):

82

$$W_{SOA} = \frac{\rho_w}{\rho_{SOA}} \frac{C_{SOA}\kappa}{\left(\frac{1}{RH} - 1\right)}$$
(1)

where ρ_w is the density of water, ρ_{SOA} the SOA density, C_{SOA} the SOA concentration, κ the SOA hygroscopicity parameter and RH the relative humidity in the 0–1 scale. More details about the ISORROPIA-lite can be found in Kakavas et al. (2022).

86

87 2.2 PMCAMx description and application

PMCAMx (Karydis et al., 2010; Tsimpidi et al., 2010) is a three dimensional 88 chemical transport model based on CAMx (Environ, 2006), which simulates 89 90 horizontal and vertical advection and dispersion, dry and wet deposition, as well as aqueous, gas, and aerosol chemistry. The mechanism used in this work for gas-phase 91 92 chemistry simulations is the Carbon Bond 05 (CB5) (Yarwood et al., 2005) and includes 190 reactions of 79 gas species. To describe the aerosol size and composition 93 94 distribution 10-size sections (from 40 nm to 40 µm) are used assuming that all 95 particles in each size bin have the same composition. Equilibrium is always assumed between the bulk aerosol and gas phases. The partitioning of semi-volatile inorganic 96 species between the gas and particulate phases is simulated by ISORROPIA-lite. 97 98 Weighting factors based on each size bin's effective surface area are used to distribute 99 to the various size bins the mass transferred between the two phases in each time step 100 (Pandis et al., 1993). For the simulation of organic aerosols, the volatility basis set approach (Donahue et al., 2006) is used. POA is simulated using eight volatility bins 101





102 (from 10^{-1} to $10^{6} \mu \text{g m}^{-3}$) at 298 K, while for SOA four volatility bins (1, 10, 10^{2} , 103 $10^{3} \mu \text{g m}^{-3}$) at 298 K are used (Murphy and Pandis, 2009). For the major point 104 sources, the NO_x plumes are simulated using the Plume-in-Grid (PiG) approach 105 (Karamchandani et al., 2011; Zakoura and Pandis, 2019).

106 We applied PMCAMx over the continental United States during 2010. The modeling domain includes northern Mexico and southern Canada and covers a $4752 \times$ 107 2952 km² region (Figure S1). The model grid consists of 10,824 cells with horizontal 108 109 dimensions of 36×36 km. The meteorological inputs were provided by the Weather 110 Research Forecasting model (WRF v3.6.1) using a horizontal resolution of 12 \times 12 km. The gaseous and primary particle emissions were developed by Xing et 111 112 al. (2013). More details about the meteorological inputs and the emissions can be 113 found in Skyllakou et al. (2021).

114 To quantify the SOAW effects on inorganic aerosol thermodynamics three 115 simulations were performed. The first was a simulation neglecting SOAW and 116 including only inorganic aerosol water. Two additional simulations were performed: 117 one where κ of SOA was assumed to be equal to 0.1 and one with κ =0.2 to examine 118 how SOA hygroscopicity affects total fine aerosol water content and PM levels and 119 composition. A SOA density of 1 g cm⁻³ was assumed in the simulations. The SOA 120 exists mostly in submicrometer particles so our subsequent study focuses on PM₁.

121

122 **3. Results**

123 3.1 Effects of SOAW on PM₁ water levels

The annual average PM_1 water ground-level concentrations neglecting SOAW are 124 shown in Figure 1. Higher PM₁ water concentrations from 8 to 18 μ g m⁻³ are 125 predicted in the north-eastern part of the US due to the higher inorganic PM_1 126 concentrations (Figure S2) and RH levels in that area. When SOAW is present in the 127 128 simulations, total PM₁ water levels increase everywhere with higher fractional increases in the south-eastern US (up to 50% when κ =0.1 and up to 100% when κ =0.2 129 in Alabama and north-western Mexico) due to higher SOA levels (Figure S3). In the 130 north-eastern US, lower fractional increases are predicted (10–15% when κ =0.1 and 131 20–30% when κ =0.2). In general, assuming a κ of SOA equal to 0.2 instead of 0.1 132 increases the corresponding amount of SOAW by about a factor of two. Figure 1 133 shows the distributions of fractional increase change in the annual PM_1 water levels at 134 ground level from SOAW. Total PM₁ water average concentrations increase from 20 135





to 30% in about 60% of the modeling domain when κ =0.1. For κ =0.2, the corresponding increase is from 40 to 60%.

Predicted SOA levels are higher during summertime (Figure S3) since the 138 139 emissions and oxidation rates of volatile organic compounds (VOCs) are higher 140 (Zhang et al., 2013; Freney et al., 2014; Skyllakou et al., 2014; Fountoukis et al., 2016). However, even during wintertime fresh biomass burning emissions exposed to 141 NO_2 and O_3 can form significant amounts of SOA in periods with low OH levels 142 143 (Kodros et al., 2020). Higher total PM1 water concentrations are predicted during 144 winter (Figure S4) since the RH levels and inorganic fine aerosol concentrations are higher; especially nitrate and chloride which increasingly partition to the aerosol 145 146 phase as temperature decreases (Guo et al., 2017). Higher fractional increases in fine aerosol water levels (up to 5 times) due to SOAW are predicted during summer in the 147 south-eastern part of US where SOA concentrations are higher. This corresponds to 148 increases to average fine aerosol water concentrations up to $8 \ \mu g \ m^{-3}$. 149

Ammonium nitrate and ammonium sulfate are the inorganic salts that contribute the most to the total PM₁ water levels (Figure S5). SOAW also contributes significantly to the total PM₁ water levels especially in the south-eastern US (about 30 and 50% of total PM₁ water when κ =0.1 and κ =0.2 respectively), when the mass fraction of SOA in dry PM₁ exceeds 30%.

155

156 **3.2 Effects of SOAW on total dry PM1 levels**

157 Higher dry PM_1 concentrations are predicted for the eastern part of the US (up to 15 $\mu g m^{-3}$) in the base case (Figure 2). These dry PM₁ levels increase slightly up to 0.6% 158 and 1.2% due to SOAW when κ =0.1 and κ =0.2 for SOA is assumed. The highest 159 annual average fractional increase in total dry PM₁ levels is predicted in California 160 (1% when κ =0.1 and 2% when κ =0.2). The probability density (Figure 2) indicates 161 162 that in about 60% of the modeling domain total dry fine aerosol concentrations increase up to 0.3% when κ =0.1. For κ =0.2, the corresponding increase is from 0.4 to 163 2%. The areas of the highest PM_1 increase correspond to regions where aerosol pH 164 tends to be relatively high (Pye et al., 2020). In these areas, nitric acid and ammonia 165 can condense and increase aerosol mass because of the increase in water from the 166 SOA. Because of this partitioning change, the predicted gas-phase concentrations of 167 semi-volatile inorganic components decreased on average when SOAW was 168 considered (Figure S6). SOAW had a negligible absolute impact on the small fine 169





- chloride concentrations in this period (Figure S2). However, in periods during which
 chloride salts and SOA contribute significantly to the total dry (e.g. during intense
 biomass burning periods), fine chloride concentrations could also change (Metzger et
- al., 2006; Fountoukis et al., 2009; Gunthe et al., 2021).
- 174

175 3.3 Effects of SOAW on PM₁ components

The annual average results indicate that SOAW mainly affects fine aerosol water 176 177 levels. To better analyze the effects of SOAW we focus on the temporal evolution of 178 the predicted levels of PM1 components in four sites (Figure S1) with different characteristics (Table S1). The presence of SOAW increased PM1 water 179 concentrations in all sites from 1% to almost an order of magnitude (Figure 3). 180 181 However, these fractional increases most of the time correspond to PM_1 water concentration increases of a few $\mu g m^{-3}$ (Figure S7) because they occur under low RH 182 levels. During higher RH periods (80 to 100%), the PM1 water levels are predicted to 183 increase up to 100 μ g m⁻³ (e.g. in Toronto). 184

185 Total dry PM1 concentrations during most of the simulated period increase on average less than 1% in all sites (Figure 3) due to SOAW. There are periods, however, 186 187 with higher fractional increases (up to 10%) and even small decreases (up to 5%) in total dry fine aerosol levels in the examined sites. The decreases can be explained 188 189 because SOAW increases the size of particles and therefore their dry deposition rate (Nenes et al., 2020). Depending on SOA hygroscopicity, increases up to 1.5 µg m⁻³ 190 for nitrate and 0.5 μ g m⁻³ for ammonium are predicted (Figure S7). Fine nitrate 191 increases of 10% were more frequent in the examined sites; however higher increases 192 193 up to 200% are predicted during the simulated period (Figure 4). As expected, higher increases can occur more often with higher assumed SOA hygroscopicity. 194

195

196 4. Discussion

197 Aerosol liquid water has a profound impact on aerosol processes, chemical 198 composition and their impacts. By including the effects of organic water on 199 inorganics thermodynamic equilibrium we show that SOAW can substantially 200 increase aerosol water levels, on an average up to 60% over the majority of the 201 domain. As a consequence, total dry PM_1 levels can also increase but the changes are 202 small (up to 2% on an annual average basis). Locally these effects can be much more





significant during periods of high RH and SOA levels (fine nitrate fractional increasescan be as high as 200%).

The effects vary with season. During summer, the RH is lower and SOA levels are higher leading to higher fractional increases in aerosol water (Figure S4) but lower absolute mass changes. During summer the fractional increases in total dry fine aerosol concentrations are lower than in wintertime (Figure S8). Responsible for the total dry fine aerosol concentration increases are nitrate and ammonium (Figure S2). These compounds partition together (as deliquesced ammonium nitrate) to the particulate phase to satisfy equilibrium due to the additional water mass of SOA.

The increases in total dry PM_1 and fine aerosol water levels depend on SOA concentrations, hygroscopicity value, RH levels and the particle phase fractions of inorganic species. The SOAW effect on aerosol water is approximately proportional to the assumed hygroscopicity parameter κ .

Aerosol liquid water directly affects the PM sensitivity and dry deposition rates, with direct implications for emissions control policy (Nenes et al., 2020; Nenes et al., 2021; Sun et al., 2021). Given this, and the important role of SOAW for climate forcing, visibility and chemistry, its inclusion in future studies is highly recommended. ISORROPIA-lite provides a simple and computationally effective approach for the simulation of SOAW.

222

Code and Data Availability. The model code and data used in this study are available
from the authors upon request (spyros@chemeng.upatras.gr and
athanasios.nenes@epfl.ch).

226

Author Contributions. SK incorporated ISORROPIA-lite in PMCAMx, carried out the
 simulations, analyzed the results and wrote the manuscript. SN and AN conceived and
 led the study and helped in the writing of the manuscript.

230

231 *Competing Interests.* The authors declare no competing financial interest.

232

Acknowledgements. This work was supported by the project FORCeS funded from
the European Union's Horizon 2020 research and innovation programme under grant
agreement No 821205, and project PyroTRACH (ERC-2016-COG) funded from





- 236 H2020-EU.1.1. Excellent Science European Research Council (ERC), project ID
- 237 726165.
- 238

239 **References**

- 240 Baker, A., Kanakidou, M., Nenes, A., Myriokefalitakis, S., Croot, P.L., Duce, A. D.,
- Gao, Y., Guieu, C., Ito, A., Jickells, T. D., Mahowald, N. M., Middag, R.,
 Perron, M. M. G., Sarin, M. M., Shelley, R., and Turner, D. R.: Changing
 atmospheric acidity as a modulator of nutrient deposition and ocean
 biogeochemistry, Sci. Adv., 7, doi: 10.1126/sciadv.abd8800, 2021.
- Bougiatioti, A., Nikolaou, P., Stavroulas, I., Kouvarakis, G., Weber, R., Nenes, A.,
 Kanakidou, M., and Mihalopoulos, N.: Particle water and pH in the eastern
 Mediterranean: source variability and implications for nutrient availability,
 Atmos. Chem. Phys., 16, 4579–4591, 2016.
- Carlton, A. G. and Turpin, B. J.: Particle partitioning potential of organic compounds
 is highest in the Eastern US and driven by anthropogenic water, Atmos. Chem.
 Phys., 13, 10203–10214, 2013.
- Donahue, N. M., Robinson, A. L., Stanier, C. O., and Pandis, S. N.: Coupled
 partitioning, dilution, and chemical aging of semivolatile organics, Environ.
 Sci. Technol., 40, 2635–2643, 2006.
- Environ: Comprehensive Air Quality Model with Extensions Version 4.40. Users
 Guide. ENVIRON Int. Corp., Novato, CA, http://www.camx.com, 2006.
- Fountoukis, C. and Nenes, A.: ISORROPIA II: a computationally efficient
 thermodynamic equilibrium model for K⁺-Ca²⁺-Mg²⁺-NH₄⁺-Na⁺-SO₄²⁻-NO₃⁻Cl⁻-H₂O aerosols, Atmos. Chem. Phys., 7, 4639–4659, 2007.
- Fountoukis, C., Nenes, A., Sullivan, A., Weber, R., Van Reken, T., Fischer, M.,
 Matías, E., Moya, M., Farmer, D., and Cohen, R. C.: Thermodynamic
 characterization of Mexico City aerosol during MILAGRO 2006, Atmos.
 Chem. Phys., 9, 2141–2156, 2009.
- Fountoukis, C., Megaritis, A. G., Skyllakou, K., Charalampidis, P. E.,
 Denier van der Gon, H. A. C., Crippa, M., Prévôt, A. S. H., Fachinger, F.,
 Wiedensohler, A., Pilinis, C., and Pandis, S. N.: Simulating the formation of
 carbonaceous aerosol in a European Megacity (Paris) during the MEGAPOLI
 summer and winter campaigns, Atmos. Chem. Phys., 16, 3727–3741, 2016.





269	Freney, E. J., Sellegri, K., Canonaco, F., Colomb, A., Borbon, A., Michoud, V.,
270	Doussin, JF., Crumeyrolle, S., Amarouche, N., Pichon, JM., Bourianne, T.,
271	Gomes, L., Prevot, A. S. H., Beekmann, M., and Schwarzenböeck, A.:
272	Characterizing the impact of urban emissions on regional aerosol particles:
273	airborne measurements during the MEGAPOLI experiment, Atmos. Chem.
274	Phys., 14, 1397–1412, 2014.
275	Gunthe, S. S., Liu, P., Panda, U., Raj, S. S., Sharma, A., Derbyshire, E., Reyes-
276	Villegas E., Allan, J., Chen, Y., Wang, X., Song, S., Pöhlker, M. L., Shi, L.,
277	Wang, Y., Kommula, S. M., Liu, T., Ravikrishna, R., McFiggans, G., Mickey,
278	L. J., Martin, S. T., Pöschl, U., Andreae, M. O., and Coe, H.: Enhanced
279	aerosol particle growth sustained by high continental chlorine emission in
280	India., Nat. Geosci., 14, 77–84, 2021.
281	Guo, H., Xu, L., Bougiatioti, A., Cerully, K. M., Capps, S. L., Hite Jr., J. R., Carlton,
282	A. G., Lee, SH., Bergin, M. H., Ng, N. L., Nenes, A., and Weber, R. J.: Fine-
283	particle water and pH in the southeastern United States, Atmos. Chem. Phys.
284	15, 5211–5228, 2015.
285	Guo, H., Liu, J., Froyd, K. D., Roberts, J. M., Veres, P. R., Hayes, P. L., Jimenez, J.
286	L., Nenes, A., and Weber, R. J.: Fine particle pH and gas-particle phase
287	partitioning of inorganic species in Pasadena, California, during the 2010
288	CalNex campaign, Atmos. Chem. Phys., 17, 5703-5719, 2017.
289	Guo, H., Li, X., Li, W., Wu, J., Wang, S., and Wei, J.: Climatic modification effects
290	on the association between PM_1 and lung cancer incidence in China, BMC
291	public health, 21, 880, 2021.
292	Jathar, S.H., Mahmud, A., Barsanti, K.C., Asher, W. E., Pankow, J. F., and Kleeman
293	M. J.: Water uptake by organic aerosol and its influence on gas/particle
294	partitioning of secondary organic aerosol in the United States, Atmos.
295	Environ., 129, 142–154, 2016.
296	Jin, X., Wang, Y., Li, Z., Zhang, F., Xu, W., Sun, Y., Fan, X., Chen, G., Wu, H., Ren,
297	J., Wang, Q., and Cribb, M.: Significant contribution of organics to aerosol
298	liquid water content in winter in Beijing, China, Atmos. Chem. Phys., 20,
299	901–914, 2020.
300	Kakavas, S., Pandis, S. N., and Nenes, A.: ISORROPIA-lite: A comprehensive
301	atmospheric aerosol thermodynamics module for Earth System Models, Tellus
302	B, 74, 1–23, 2022.





303	Karamchandani, P., Vijayaraghavan, K., and Yarwood, G.: Sub-grid scale plume
304	modeling, Atmosphere, 2, 389-406, 2011.
305	Karydis, V. A., Tsimpidi, A. P., Fountoukis, C., Nenes, A., Zavala, M., Lei, W.,
306	Molina, L. T., and Pandis, S. N.: Simulating the fine and coarse inorganic
307	particulate matter concentrations in a polluted megacity, Atmos. Environ., 44,
308	608–620, 2010.
309	Kodros, J. K., Papanastasiou, D. K., Paglione, M., Masiol, M., Squizzato, S., Florou,
310	K., Skyllakou, K., Kaltsonoudis, C., Nenes, A., and Pandis, S. N.: Rapid Dark
311	Aging of Biomass Burning as an Overlooked Source of Oxidized Organic
312	Aerosol., Proc. Natl. Acad. Sci. U.S.A., 117, 33028-33033, 2020.
313	Koehler, K. A., Kreidenweis, S. M., DeMott, P. J., Petters, M. D., Prenni, A. J., and
314	Carrico, C. M.: Hygroscopicity and cloud droplet activation of mineral dust
315	aerosol, Geoph. Res. Let., 36, (8), 2009.
316	Liao, H. and Seinfeld, J. H.: Global impacts of gas-phase chemistry aerosol
317	interactions on direct radiative forcing by anthropogenic aerosols and ozone, J.
318	Geophys. Res., 110, D18208, 2005.
319	Metzger, S., Mihalopoulos, N., and Lelieveld, J.: Importance of mineral cations and
320	organics in gas-aerosol partitioning of reactive nitrogen compounds: case
321	study based on MINOS results., Atmos. Chem. Phys., 6, 2549–2567, 2006.
322	Murphy, B. N. and Pandis, S. N.: Exploring summertime organic aerosol formation in
323	the Eastern United States using a regional-scale budget approach and ambient
324	measurements, J. Geophys. Res., 115, (D24216), 2010.
325	Nenes, A., Pandis, S. N., Weber, R. J., and Russell, A.: Aerosol pH and liquid water
326	content determine when particulate matter is sensitive to ammonia and nitrate
327	availability, Atmos. Chem. Phys., 20, 3249-3258, 2020.
328	Nenes, A., Pandis, S. N., Kanakidou, M., Russell, A. G., Song, S., Vasilakos, P., and
329	Weber, R. J.: Aerosol acidity and liquid water content regulate the dry
330	deposition of inorganic reactive nitrogen, Atmos. Chem. Phys., 21, 6023-
331	6033, 2021.
332	Nguyen, T. K. V., Zhang, Q., Jimenez, J. L., Pike, M., and Carlton, A. G.: Liquid
333	water: ubiquitous contributor to aerosol mass. Environ. Sci. Tech. Let., 3,
334	257–263, 2016.





335	Pandis, S. N., Wexler, A. S., and Seinfeld, J. H.: Secondary organic aerosol formation
336	and transport - II. Predicting the ambient secondary organic aerosol size
337	distribution, Atmos. Environ., 27, 2403–2416, 1993.
338	Petters, M. D. and Kreidenweis, S. M.: A single parameter representation of
339	hygroscopic growth and cloud condensation nucleus activity, Atmos. Chem.
340	Phys., 7, 1961–1971, 2007.
341	Pye, H. O. T., Nenes, A., Alexander, B., Ault, A. P., Barth, M. C., Clegg, S. L.,
342	Collett Jr, J. L., Fahey, K. M., Hennigan, C. J., Herrmann, H., Kanakidou, M.,
343	Kelly, J. T., Ku, IT., McNeill, V. F., Riemer, N., Schaefer, T., Shi, G.,
344	Tilgner, A., Walker, J. T., Wang, T., Weber, R., Xing, J., Zaveri, R. A., and
345	Zuend, A.: The acidity of atmospheric particles and clouds, Atmos. Chem.
346	Phys., 20, 4809–4888, 2020.
347	Seinfeld, J. H. and Pandis, S. N.: Atmospheric chemistry and physics: From air
348	pollution to climate change, Wiley: New York, 2006.
349	Skyllakou, K., Murphy, B. N., Megaritis, A. G., Fountoukis, C., and Pandis, S. N.:
350	Contributions of local and regional sources to fine PM in the megacity of
351	Paris, Atmos. Chem. Phys., 14, 2343-2352, 2014.
352	Skyllakou, K., Rivera, P. G., Dinkelacker, B., Karnezi, E., Kioutsioukis, I.,
353	Hernandez, C., Adams, P. J., and Pandis, S. N.: Changes in
354	$PM_{2.5}$ concentrations and their sources in the US from 1990 to 2010, Atmos.
355	Chem. Phys., 21, 17115–17132, 2021.
356	Song, S., Gao, M., Xu, W., Shao, J., Shi, G., Wang, S., Wang, Y., Sun, Y., and
357	McElroy, M. B.: Fine-particle pH for Beijing winter haze as inferred from
358	different thermodynamic equilibrium models, Atmos. Chem. Phys., 18, 7423-
359	7438, 2018.
360	Sun, X., Ivey, C. E., Baker, K. R., Nenes, A., Lareau, N. P., and Holmes, H. A.:
361	Confronting Uncertainties of Simulated Air Pollution Concentrations during
362	Persistent Cold Air Pool Events in the Salt Lake Valley, Utah., Environ. Sci.
363	Technol., 55, 15072–15081, 2021.
364	Tsimpidi, A. P., Karydis, V. A., Zavala, M., Lei, W., Molina, L., Ulbrich, I. M.,
365	Jimenez, J. L., and Pandis, S. N.: Evaluation of the volatility basis-set
366	approach for the simulation of organic aerosol formation in the Mexico City
367	metropolitan area, Atmos. Chem. Phys., 10, 525-546, 2010.





368	Xing, J., Pleim, J., Mathur, R., Pouliot, G., Hogrefe, C., Gan, CM., and Wei, C.:
369	Historical gaseous and primary aerosol emissions in the United States from
370	1990 to 2010, Atmos. Chem. Phys., 13, 7531-7549, 2013.
371	Yarwood, G., Rao, S., Yocke, M., and Whitten, G. Z.: Updates to the Carbon Bond
372	Chemical Mechanism: CB05, Research Triangle Park, https://camx-
373	wp.azurewebsites.net/Files/CB05_Final_Report_120805.pdf, 2005.
374	Zakoura, M. and Pandis, S. N.: Improving fine aerosol nitrate predictions using a
375	Plume-in-Grid modeling approach, Atmos. Environ., 215, 116887, 2019.
376	Zhang, Q. J., Beekmann, M., Drewnick, F., Freutel, F., Schneider, J., Crippa, M.,
377	Prevot, A. S. H., Baltensperger, U., Poulain, L., Wiedensohler, A., Sciare, J.,
378	Gros, V., Borbon, A., Colomb, A., Michoud, V., Doussin, JF., Denier van
379	der Gon, H. A. C., Haeffelin, M., Dupont, JC., Siour, G., Petetin, H.,
380	Bessagnet, B., Pandis, S. N., Hodzic, A., Sanchez, O., Honoré, C., and
381	Perrussel, O.: Formation of organic aerosol in the Paris region during the
382	MEGAPOLI summer campaign: evaluation of the volatility basis-set approach
383	within the CHIMERE model, Atmos. Chem. Phys., 13, 5767-5790, 2013.
384	Zhang, Z., Gao, J., Engling, G., Tao, J., Chai, F., Zhang, L., Zhang, R., Sang, X.,
385	Chan, C., Lin, Z., and Cao, J.: Characteristics and applications of size-
386	segregated biomass burning tracers in China's Pearl River Delta region,
387	Atmos. Environ., 102, 290-301, 2015.
388	
389	
390	
391	
392	
393	
205	
396	
397	
398	
399	
400	
401	
402	
403	







Figure 1. Maps of: (a) annual average PM_1 water ground-level concentrations neglecting SOAW, (b) annual average fractional increase of PM_1 water when SOAW is present in the simulations with κ =0.1 and, (c) with κ =0.2 during 2010. The probability density as a function of fractional increase in the annual PM_1 water concentrations due to SOAW when: (d) κ =0.1 and (e) κ =0.2 is shown.







Figure 2. Maps of: (a) annual average total dry PM₁ ground-level concentrations neglecting SOAW, (b) annual average fractional increase of total dry PM₁ when SOAW is present in the simulations with κ =0.1 and, (c) with κ =0.2 during 2010. The probability density as a function of fractional increase in the annual total dry PM₁ concentrations due to SOAW when: (d) κ =0.1 and (e) κ =0.2 is shown.









Figure 3. Box plots for fractional change in the hourly: (a) PM₁ water and (b) total dry PM₁ due to SOAW when κ =0.1 and κ =0.2 for Sacramento, California; Houston, Texas; Atlanta, Georgia; and Toronto, Canada during 2010. The red line represents the median, the black dot is the mean value, the upper box line is the upper quartile (75%) and the lower box line is the lower quartile (25%) of the distribution. A negative change corresponds to a decrease.

- 479
- 480
- 481
- 482
- 483 484
- 485







486 487

Figure 4. The probability density as a function of fractional increase in the hourly PM₁ nitrate due to SOAW when κ =0.1 and κ =0.2 for: (a) Sacramento, California; (b) Houston, Texas; (c) Atlanta, Georgia; and (d) Toronto, Canada during 2010.