



Supplement of

Significant source of secondary aerosol: formation from gasoline evaporative emissions in the presence of SO_2 and NH_3

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15 S1. AMS Corrections: Comparison with SMPS Measurements

16 Theoretically, the sum of the secondary aerosol (SA) mass measured by HR-ToF-AMS should be equal to the mass calculated from the SMPS size distributions. However, both methods have their 17 limitations, in which SMPS measures particle mobility diameter, while HR-ToF-AMS measures mass. 18 19 Therefore, particle shape and density must be assumed before converting SMPS measurements to mass. 20 Here, we assume that particles are spherical, and the density of SA were calculated from the equation 21 $\rho = d_{va}/d_m$, where d_{va} is the mean vacuum aerodynamic diameter measured by an HR-ToF-AMS and d_m is the mean electrical mobility diameter measured by SMPS (DeCarlo et al., 2004). However, fractal-22 23 like particles will cause the SMPS to overestimate the spherical equivalent diameter and therefore 24 overestimate the particle mass. While, HR-ToF-AMS tends to underestimate the SA mass due to the 25 transmission efficiency (Liu et al., 2007) and collection efficiency (Takegawa et al., 2005). 26 For all the experiments with the discrepancies between HR-ToF-AMS and SMPS, we assume that 27 the difference in mass has the same chemical composition as the measured chemical species (i.e., 28 organics, nitrate, sulfate, and ammonium). And then a scaling factor (AMS_{sf}) was calculated for each 29 experiment to correct the SA mass measured by HR-ToF-AMS and close the gap with the SMPS

30 measurement. The scaling factor could be calculated as following equation:

$$AMS_{sf} = \frac{C_{SMPS}}{C_{Org} + C_{NO_3} + C_{SO_4} + C_{NH}}$$

in which C_{SMPS} is the SA mass concentration derived from SMPS, C_{Org} , C_{NO_3} , C_{SO_4} and C_{NH_4} are the mass concentrations of organics, nitrate, sulfate, and ammonium measured by HR-ToF-AMS, respectively. The AMS_{sf} for each time step after wall loss correction is calculated and used to scale the AMS data for the entire experiment. For all the experiments the average AMS_{sf} ranged from 1.09 to 36 1.23.

37 S2. Vapor wall loss and gas-particle partitioning timescales

38 The loss of vapor by condensation onto the wall is generally considered to be a first-order process,

39 which can be characterized by the first-order wall-loss coefficient k_w (s⁻¹). According to the following

40 equation reported by McMurry and Grosjean (1985), the value of k_w is equal to:

41
$$k_{\rm w} = \frac{A}{V} \times \frac{\alpha_{\rm w} \overline{c}}{1.0 + \frac{\pi}{2} \times \left[\frac{\alpha_{\rm w} \overline{c}}{4(k_{\rm e} D_{\rm gas})^{0.5}}\right]}$$

42 in which A and V are the surface and volume of the smog chamber, respectively. For our cuboid smog 43 chamber ($L \times W \times H = 3.0 \times 2.5 \times 4.0 \text{ m}$), A=59 m², V=30 m³. α_w is the mass accommodation coefficient 44 of vapors onto the chamber walls, \bar{c} is the mean thermal speed of the molecules, k_e is the coefficient of 45 eddy diffusion, and D_{gas} is the gas-phase diffusivity.

For a given vapor molecule, the mean thermal speed \overline{c} could be calculated according to the following equation:

48
$$\overline{c} = \sqrt{\frac{8RT}{\pi MW}}$$

in which R is the ideal gas constant (i.e., 8.314 J mol⁻¹ K⁻¹), T is the experimental temperature (T=299.15
K in this study), and MW is the molecular weight (an upper bound and a lower bound of molecular mass
of organic vapors was adopted, i.e., 100 g mol⁻¹ and 300 g mol⁻¹ in this study).
D_{gas} is assumed to vary with molecular weight (MW) and is equal to D_{CO2}(MW_{CO2}/MW), with

53 $D_{CO_2} = 1.38 \times 10^{-5} \text{ m}^2 \text{ s}^{-1}$. Therefore, this leaves k_e and α_w as the two key unknowns. For the value of k_e,

 0.015 s^{-1} was estimated according to the values reported by previous studies for a 28 m³ Caltech chamber

55 (Loza et al., 2012; McMurry and Rader, 1985; Zhang et al., 2014). For the value of α_w , 10⁻⁵ was adopted

according to the experimental results of Matsunaga and Ziemann (2010).

57 Then the timescale associated with vapor-wall loss (τ_{g-w}) is calculated to be:

58
$$\tau_{g-w} = k_w^{-1}$$

59 for the timescale associated with reaching gas-to-particle partitioning equilibrium ($\overline{\tau}_{g-p}$), which

60 varies with particle number concentration and diameter, and could be approximately calculated to be:

$$\overline{\tau}_{g-p} = (2\pi N_p \overline{D_p} D_{gas} \overline{F_{FS}})^{-1}$$

62 in which N_p is the particle number concentration, $\overline{D_p}$ is the particle mean diameter, D_{gas} is the gas-63 phase diffusivity and $\overline{F_{FS}}$ is the correction to the mass transfer flux due to noncontinuum effects and 64 imperfect accommodation given in the following equation: $0.75\alpha(1 \pm k_{-})$

65
$$\overline{F_{FS}} = \frac{0.75\alpha(1+k_n)}{k_n^2 + k_n + 0.283k_n\alpha + 0.75\alpha}$$

in which α is the mass accommodation coefficient onto particles, for which the value of 0.002 was

adopted in this study, and k_n is the Knudsen number, defined as:

$$k_n = \lambda / R_p$$

in which λ is the gas mean free path, which could be calculated as following equation:

$$\lambda = \frac{3D_{gas}}{\overline{c}}$$

In our study, the SA yields were underestimated by a factor of 1.97–2.82 fold when considering the ratio of these two timescales (i.e., $\bar{\tau}_{g-p}/\tau_{g-w}$), which showed a decreasing trend with increasing SO₂ and NH₃ initial concentrations, suggesting that an increasing proportion of vapors is partitioned onto the suspended particle surface rather than the chamber wall. Meanwhile, the wall loss of sulfuric acid gas was also considered using this ratio (i.e., $\bar{\tau}_{g-p}/\tau_{g-w}$) to correct the sink of sulfur species.

76 S3. Positive matrix factorization (PMF)

77	Positive Matrix Factorization (PMF) (Paatero, 1997; Paatero and Tapper, 1994) is a receptor model
78	and multivariate factor analysis tool that decomposes a matrix of speciated sample data into two matrices,
79	namely factor contributions and factor profiles. In recent years, the PMF model was used for the analysis
80	of high-resolution (HR) mass spectra data which can provide better separation of different organic
81	components (Liu et al., 2014). This model was expressed as a bilinear factor model, namely, $x_{ij}=\sum_{p}g_{ip}f_{pj}$
82	$+ e_{ij}$, where i and j refer to values of j species in i samples, respectively, p is the number of factors in the
83	solution, and used a least-squares fitting process, minimizing a quality of fit parameter. In our study, we
84	used the PMF software together with a modified version of the CU AMS PMF Execute Calcs Tool v
85	2.06, which was developed by Ulbrich et al. (2009), to analyze the HR mass spectra (m/z 12 – 170)
86	(Zhang et al., 2011). The data and noise matrices input into the PMF analysis were generated from the
87	PIKA version 1.15D. Ions were classified and down-weighted according to the signal-to-noise ratios
88	(SNR). 0.2 <snr<2 2,="" a="" and="" as="" by="" classified="" down-weighted="" factor="" ions="" of="" snr<0.2="" td="" the="" was="" was<="" weak=""></snr<2>
89	bad ions and removed from the analysis and noise values of CO_2^+ -related peaks at m/z 16 (O), 17 (HO),
90	18 (H ₂ O), 28 (CO), and 44 (CO ₂) were down-weighted.

Commente	Volume Fraction	Compounds	Volume Fraction
Compounds	(v/v %)		(v/v %)
1,3-Butadiene	10.67	n-Octane	1.71
1-Pentene	0.00	p-Xylene and m-Xylene	2.28
trans-2-Pentene	0.06	Ethylbenzene	1.82
cis-2-Pentene	0.06	Nonane	0.75
Isoprene	1.33	o-Xylene	1.09
2,2-Dimethylbutane	1.11	Styrene	0.00
2,3-Dimethylbutane	7.74	Isopropylbenzene	0.30
2-Methylpentane	7.83	n-Propylbenzene	2.11
3-Methylpentane	5.63	1,3,5-Trimethylbenzene	0.93
1-Hexene	0.00	m-Ethyltoluene	0.93
n-Hexane	9.89	p-Ethyltoluene	0.93
2,4-Dimethylpentane	2.42	n-Decane	0.00
Methylcyclopentane	3.70	o-Ethyltoluene	0.60
Cyclohexane	1.91	1,2,4-Trimethylbenzene	5.12
2-Methylhexane	2.18	1,2,3-Trimethylbenzene	0.92
3-Methylhexane	2.62	m-Diethylbenzene	0.17
2,3-Dimethylpentane	2.53	p-Diethylbenzene	0.17
Benzene	0.58	n-Undecane	0.00
2,2,4-Trimethylpentane	3.87	n-Dodecane	0.83
n-Heptane	5.12		
Methylcyclohexane	2.43		
2,3,4-Trimethylpentane	1.20		
2-Methylheptane	0.62		
3-Methylheptane	0.59		
Toluene	4.90		

Table S1. The volume fraction of detected compounds for gasoline utilized in this study.



94 Fig. S1. Schematic of the RCEES-CAS smog chamber facility. E: Electromagnetic valve; T: Three-way valve. M:

95 Mass flow controller.



Fig. S2. Time variations of inorganic gas-phase species (a) NO, (b) NO_x, (c) NO_x-NO, (d) O₃, and (e) SO₂ in photooxidation of gasoline/NO_x in the presence or absence of SO₂ and NH₃. Letters in abbreviations represent the reactants
introduced into the chamber reactor, i.e., "G" represents gasoline, "N" represents nitrogen oxides, "S" represents
sulfur dioxide, "A" represents ammonia.



Fig. S3. Time variations of organic gas-phase species (a) Benzene, (b) Toluene, (c) C2-Benzene, (d) C3-Benzene,
(e) C4-Benzene, (f) Methylcyclopentane, and (g) Methylcyclohexane in photo-oxidation of gasoline/NO_x in the
presence or absence of SO₂ and NH₃. Letters in abbreviations represent the reactants introduced into the chamber
reactor, i.e., "G" represents gasoline, "N" represents nitrogen oxides, "S" represents sulfur dioxide, "A" represents
ammonia.



108 Fig. S4. Time variations of (a) Benzene and (b) Toluene measured by PTR-TOF and GC-MS during a typical chamber

109 experiment (experiment GN).



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111 Fig. S5. Time variations of acetic acid during the photo-oxidation of gasoline/NO_x in the presence or absence of SO₂ and NH₃.

- 112 Letters in abbreviations represent the reactants introduced into the chamber reactor, i.e., "G" represents gasoline, "N"
- 113 represents nitrogen oxides, "S" represents sulfur dioxide, "A" represents ammonia.



Fig. S6. Time series of the smaller size distributions (4–160 nm) derived from SMPS equipped with a nanometer differential
 mobility analyzer (Nano-DMA) for the generated secondary aerosol during the photo-oxidation experiments with different





Fig. S7. Fitted peaks of average W-mode mass spectrum of organosulfur compounds (OS), (a) CSO⁺, (b) CH₃SO_{2⁺}, (c)
CH₃SO_{3⁺}.



122 Fig. S8. Time series of the ammonium aerosol formed during the photo-oxidation experiments with different SO_2







126

127 Fig. S9. (a) Linear relationship between the concentration of chemical species and NH₃ under different NH₃ initial 128 concentration conditions. Each line (green (organic), blue (nitrate), red (sulfate), and orange (ammonium)) represents a linear 129 fitting and the k values are the corresponding slopes for each chemical species. (b) Time series of the size distributions (20-700 130 nm) for the secondary aerosol generated during the photo-oxidation experiments with different NH₃ concentrations. D_{p,max} and $N_{\rm max}$ represent the maximal diameter and number concentration of generated secondary aerosol, respectively, during each 131 132 photo-oxidation experiment. (c) Time series of the smaller size distributions (4-160 nm) derived from SMPS equipped with a 133 nanometer differential mobility analyzer (Nano-DMA) for the generated secondary aerosol during the photo-oxidation 134 experiments with different NH3 concentrations.



136 Fig. S10. Average O/C and H/C in SOA formed from the photo-oxidation of gasoline vapor at different concentrations of NH₃





135



140 Fig. S11. Mass spectra of the two factors identified from the PMF analysis to the AMS data derived from the experiments at

different concentrations of (a) SO₂ (Exp. GN, SGN1, SGN2, SGN3 and SGN4) and (b) NH₃ (Exp. GN, AGN1 and AGN2).



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143 Fig. S12. Average O/C and H/C in SOA formed from the photo-oxidation of gasoline vapor at different concentrations of SO₂

144 (Exp. GN, SGN1, SGN2, SGN3 and SGN4).





146 Fig. S13. Linear relationship between the concentration of SO₂ (or NH₃) and the SA yield.

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148

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