- 1 Strong light scattering of highly oxygenated organic aerosols impacts
- 2 significantly on visibility degradation
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33 Table of contents

34	1.	Site information
35		Figure S1. Site environments and the container (green one)
36	2. Q	-ACSM analysis
37		Figure S2. Diagnostic plots of the 4-factor solution in the unconstrained PMF
38		PMF results7
39		Figure S3. Mass spectra, diurnal variations and time series of 3-factor solution from
40		unconstrained PMF7
41		Figure S4. Mass spectra, diurnal variations, and time series of 4-factor solution from
42		unconstrained PMF8
43		Figure S5. Mass spectra, diurnal variations, and time series of 5-factor solution from
44		unconstrained PMF9
45		Figure S6. Diagnostic plots of the 5-factor solution in the unconstrained PMF during the spring
46		festival period from 11th to 25th February 202110
47		Figure S7. Mass spectra, diurnal variations, and time series of 5-factor solution from
48		unconstrained PMF during the spring festival period from 11th to 25th February 202111
49		ME-2 results12
50		Figure S8. Mass spectra, diurnal variations, and time series of ME-2(a-value=0.1) under
51		the 4-factor solution12
52		Figure S9. Mass spectra, diurnal variations, and time series of ME-2(a-value=0.2) under
53		the 4-factor solution
54		Figure S10. Mass spectra, diurnal variations, and time series of ME-2(a-value=0.3) under
55		the 4-factor solution14
56	3. I	Discussions on traditional multilinear regression model14
57		Table S1. Square of correlation coefficients between aerosol components
58		Table S2. Square of correlation coefficients between changes of aerosol components for
59		identified cases16
60	4. V	/isibility contributions estimation16
61	5. (Other Figures
62		Figure S11. Simulated relationships between VSE _{PM1} and Rsca using Mie theory through
63		varying volume geometric mean D_{gv} of lognormal size distributions from 100 to 700 nm under
64		different standard deviation (σg) conditions
65		Figure S12. Histogram of ambient relative humidity (RH) during the observation period20
66		Figure S13. Aerosol light scattering enhancement measurements (fRH) at 525 nm from 13 th to
67		26 th February with RH range of 60-90%
68	Ref	ferences
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73 **1. Site information**

The observation site locates in Haizhu wetland park, which is surrounded by roads, business, and residential districts, however, at least ~1 km away from the observation site.



- 77 Figure S1. Site environments and the container (green one).
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81 2.Q-ACSM analysis

In this study, organic aerosol (OA) spectra measured by the Q-ACSM were deconvolved into OA factors using an improved source apportionment technique called Multilinear Engine (ME-2) which is an upgrade of widely used Positive Matrix Factorization (PMF) technique and runs on a IGOR-based interface ¹. Different with traditional PMF, ME-2 offers a coefficient called a-value to constrain the spectra

variation extent of OA factor with given priori mass spectra ^{1,2}. The unconstrained runs 87 with PMF technique were firstly performed with possible factor number of 2-8. It was 88 found that four factors solution splits clearly OA factors, with solutions of 3 or 5 factors 89 show less or over split features. Results for factor number determination were shown in 90 Fig.S2-5. For example, three factors solution does not split two major primary OA 91 factors of cooking-like OA (COA) and hydrocarbon-like OA (HOA) in urban area, and 92 five factor solutions over split the oxygenated organic aerosol, thus four factors solution 93 94 was finally determined as the best. However, previous studies demonstrate that PMF solution sometimes failed in clean separating OA factors ^{1, 2}. Similar case was also 95 found in the unconstrained solution as shown in Fig.S4 that the factor 1 showed obvious 96 cooking-like primary OA (COA) features (for example high correlations with m/z 55, 97 and obvious noon peak), however showing higher oxidation feature than previously 98 reported results of COA with exceptionally high m/z 44 fraction², thus the solution has 99 defects. The a-value approach of ME-2 techniques provides additional constrains on 100 factors through introducing user defined external factor mass spectra profile, however 101 102 a priori mass spectra of COA for Q-ACSM measurements in Guangzhou urban area is lacking. Chinese spring festival (area shaded with pink color in Figure 1 of the 103 manuscript) was during the observation period and stay home policy was recommended 104 by Chinese government due to the COVID-19 epidemic, thus very small traffic flow 105 however might even higher cooking activities than usual due to the festival celebration. 106 Results of Guo, et al. $(2020)^2$ demonstrate that COA usual contribute even slightly 107 higher than HOA (Hydrocarbon-like OA), suggesting that the dominant contribution of 108 COA to primary OA during the special "spring festival and COVID epidemic stay home" 109 110 period, thus provide us a unique opportunity to identify spectra profile that most close to realistic COA spectra. The unconstrained PMF technique performed specific to the 111 spring festival period from 11th to 25th February 2021, and five factor solution with most 112 prominent COA features was determined (Fig.S6-7) although might over spilt the 113 oxygenated OA. The factor with obvious COA feature was chosen as the used defined 114 external spectra in ME-2 of four factor solutions with a values range from 0.1 to 0.5. 115

- 116 The ME-2 solution with a value of 0.2 was adopted based on correlation coefficients
- 117 with external tracers, and solutions of a values ranging from 0.1 to 0.3 as well as their
- 118 correlation coefficients with external tracers are shown in Fig.S8-10. Compared with
- results of the unconstrainted PMF, correlations of COA factor with m/z 55 has improved
- substantially (\mathbb{R}^2 increased from 0.49 to 0.77), and the determined COA factor has much
- 121 better COA features and lower O/C 3 . Note the O/C value of factors were estimated
- using the empirical relationship between f_{44} and O/C proposed by Aiken, et al. (2008)⁴.

124



127 Figure S2. Diagnostic plots of the 4-factor solution in the unconstrained PMF.





Figure S3. Mass spectra, diurnal variations and time series of 3-factor solution from
 unconstrained PMF.





156 Figure S4. Mass spectra, diurnal variations, and time series of 4-factor solution from157 unconstrained PMF.



Figure S5. Mass spectra, diurnal variations, and time series of 5-factor solution from
unconstrained PMF.



Figure S6. Diagnostic plots of the 5-factor solution in the unconstrained PMF during
the spring festival period from 11th to 25th February 2021.



Figure S7. Mass spectra, diurnal variations, and time series of 5-factor solution from unconstrained PMF during the spring festival period from 11th to 25th February 2021.

226 ME-2 results



Figure S8. Mass spectra, diurnal variations, and time series of ME-2(a-value=0.1) under the 4-factor solution.



Figure S9. Mass spectra, diurnal variations, and time series of ME-2(a-value=0.2) under the 4-factor solution.



Figure S10. Mass spectra, diurnal variations, and time series of ME-2(a-value=0.3) under the 4-factor solution.

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277 **3. Discussions on traditional multilinear regression model**

In this study, we tried to perform the traditional multiple linear regression analysis 278 with HOA, COA, LOOA, MOOA, AS (ammonium sulfate), AN (ammonium nitrate) 279 and BC as input variables and aerosol scattering coefficients as target variable. Note 280 281 that ammonium nitrate (AN) and ammonium sulfate (AS) were determined as the 282 dominant form of ammonium, and paired spot of ammonium bisulfate was treated as AS in the multiple linear regression model due to their similarity in scattering abilities. 283 Negative MSELOOA were obtained if MSE values were not constrained, and MSEs of 284 285 some aerosol components deviated significantly from previously reported ranges. If MSEs of aerosol components were constrained as positive, then an MSELOOA of zero 286

would be obtained. These results demonstrate that the multiple regression model failed 287 in retrieving aerosol MSE, and two reasons might be responsible for this failure. The 288 289 first one is mathematically fundamental, the application of multiple linear regression model perform best with independent input variables, however, the square of correlation 290 coefficients (R²) between several variables were higher than 0.5 for datasets of this 291 study (Table.S2). For example, the square of correlation coefficients between HOA and 292 LOOA, between LOOA and AN, and between HOA and BC are 0.6, 0.54 and 0.78 293 294 respectively. The second reason is associated with the observations that aerosol scattering of entire aerosol populations of PM2.5 were measured however part of the 295 aerosol mass such as PM₁ dust were not identified by the mass spectrometer ^{5, 6} and the 296 contribution of unidentified part might varies substantially ⁷. In addition, aerosol 297 scattering coefficients of PM2.5 were measured whereas mass concentrations of PM1 298 299 were quantified.

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304	Table S1. Square of	correlation coefficients	between aerosol	components
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HOA 0.37 0.6 0.11 0.79 0.16 0. COA 0.42 0.1 0.43 0.15 0. LOOA 0.37 0.57 0.43 0. MOOA 0.12 0.59 0. BC 0.22 0. 0.	0.37 0.6 0.11 0.79 0.16 0.39 0.42 0.1 0.43 0.15 0.17 0.37 0.57 0.43 0.54 0.12 0.59 0.49 0.22 0.33 0.22 0.28 0.28 0.28		COA	LOOA	MOOA	BC	AS	AN
COA 0.42 0.1 0.43 0.15 0. LOOA 0.37 0.57 0.43 0. MOOA 0.12 0.59 0. BC 0.22 0.	0.42 0.1 0.43 0.15 0.17 0.37 0.57 0.43 0.54 0.12 0.59 0.49 0.22 0.33 0.28	НОА	0.37	0.6	0.11	0.79	0.16	0.39
LOOA 0.37 0.57 0.43 0. MOOA 0.12 0.59 0. BC 0.22 0.	0.37 0.57 0.43 0.54 0.12 0.59 0.49 0.22 0.33 0.28	СОА		0.42	0.1	0.43	0.15	0.17
MOOA 0.12 0.59 0. BC 0.22 0.	0.12 0.59 0.49 0.22 0.33 0.28	LOOA			0.37	0.57	0.43	0.54
BC 0.22 0.	0.22 0.33 0.28	MOOA				0.12	0.59	0.49
46 0	0.28	BC					0.22	0.33
A5 U.		AS						0.28

314 Table S2. Square of correlation coefficients between changes of aerosol components for identified

315 cases

		∆COA	∆L00A	∆MOOA	∆BC	∆AS	∆AN
	∆HOA	0.34	0.44	0.43	0.88	0.01	0.02
	∆COA		0.34	0.56	0.29	0.05	0.07
	∆L00A			0.38	0.5	0.00	0.03
					0.4	0.12	0.07
	ΔBC					0.02	0.01
	ΔAS						0.39
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33 34							

335 **4. Visibility contributions estimation**

Based on the Koschmieder theory, atmospheric visibility is determined by atmospheric extinction coefficient σ_{ex} ⁸:

338 Visibility
$$= \frac{K}{\sigma_{ex}}$$
. Eq. S1

Where K is the Koschmieder constant, and a value of 3.0 is usually used for Asian people and thus also visibility meter ⁹. The σ_{ex} is the total atmospheric light extinction coefficient at 550 nm caused by aerosols and air molecules and can be calculated through the sum of its scattering and absorption components:

343
$$\sigma_{ex} = \sigma_{sp} + \sigma_{abs} + \sigma_{air} + \sigma_{NO_2},$$
 Eq. S2

where σ_{sp} and σ_{abs} are the aerosol scattering and absorption coefficients, σ_{air} is 344 the Rayleigh scattering by air molecules and σ_{NO_2} the absorption by NO₂ molecules. 345 Rayleigh scattering of air molecules at 550 nm under standard atmospheric pressure is 346 about 13 Mm^{-1 10}. The NO₂ absorption at 550 nm is calculated using $\sigma_{NO_2} = 0.33$. 347 $[NO_2]$, where $[NO_2]$ represents the NO₂ volume mixing ratio in units of ppb, and unit of 348 calculated σ_{NO_2} is Mm⁻¹. Aerosol absorptions at 520 and 590 nm measured by the 349 AE33 aethalometer were used to calculate aerosol absorptions (σ_{abs}) at 550 nm through 350 absorption Ångström law. 351

352 As to the aerosol scattering σ_{sp} at 550 nm, it can be calculated as based on 353 analysis of Xu, et al. $(2020)^9$:

354
$$\sigma_p = \sigma_{sp,fine} + \sigma_{sp,coarse} = \sigma_{sp,PM_{2.5}}(RH) + 0.036 \cdot \sigma_{sp,PM_{2.5}(dry)}$$
 Eq. S3

Where direct measurements of $\sigma_{sp,PM_{2.5}(dry)}$ at 525 nm were converted to $\sigma_{sp,PM_{2.5}(dry)}$ at 550 nm using measured between scattering Ångström exponent by the nephelometer. The $\sigma_{sp,PM_{2.5}}(RH)$ values at 525 nm were firstly calculated as the summation of aerosol scattering coefficients of MOOA, LOOA, HOA, COA, BC, AN and AS under ambient RH conditions by considering MSE values derived at 525 nm and aerosol hygroscopicity:

361
$$\sigma_{sp,PM_{2.5}}(RH) = \sigma_{sp,MOOA,PM_{2.5}}(RH + \sigma_{sp,LOOA,PM_{2.5}}(RH) + \sigma_{sp,HOA,PM_{2.5}}(RH) +$$

362 $\sigma_{sp,COA,PM_{2.5}}(RH) + \sigma_{sp,BC,PM_{2.5}}(RH) + \sigma_{sp,AS,PM_{2.5}}(RH) + \sigma_{sp,AN,PM_{2.5}}(RH)$ Eq. S4

363 COA, BC, HOA are hydrophobic with hygroscopic parameter κ of zero. Thus, their 364 scattering didn't change with ambient RH and are same with their values in dry state.

In addition, as discussed in Sect.4.2 of the manuscript, most HOA, COA and BC mass
reside in PM1. Thus,
$$\sigma_{sp,HOA,PM_{25}}(RH) = [HOA]_{PM1} \times MSE_{HOA,PM1}(dry)$$
,
 $\sigma_{sp,COA,PM_{25}}(RH) = [COA]_{PM1} \times MSE_{cOA,PM1}(dry)$, and $\sigma_{sp,BC,PM_{25}}(RH) =$
 $[BC]_{PM1} \times MSE_{BC,PM1}(dry)$ where [X] represents mass concentrations of aerosol
components [X]. For hydrophilic aerosol components, MOOA, LOOA, AS, and AN,
their scattering under ambient RH conditions were calculated using
 $\sigma_{sp,MOOA,PM_{25}}(RH) = [MOOA]_{PM1} \times MSE_{MOOA}^* \times \kappa_{MOOA} \times R_{sca,MOOA}$,
 $\sigma_{sp,LOOA,PM_{25}}(RH) = [MOOA]_{PM1} \times MSE_{LOOA}^* \times \kappa_{LOOA} \times R_{sca,MOOA}$,
 $\sigma_{sp,AS,PM_{25}}(RH) = [AS]_{PM1} \times MSE_{AS}^* \times \kappa_{AS}(RH) \times R_{sca,AS}$, and $\sigma_{sp,AN,PM_{25}}(RH) =$
 $[AN]_{PM1} \times MSE_{AN}^* \times \kappa_{AN}(RH) \times R_{sca,AN}$. As discussed in Sect.4.4 of the manuscript,
 $R_{sca,LOOA}$ is 0.87, and 0.63 for $R_{sca,MOOA}$, $R_{sca,AS}$ and $R_{sca,AN}$. MSE_{X}^* defined as
 $MSE_{X}^* = \frac{\sigma_{apszz}(2M_{2z})}{|X|(PM_{1})|}$ for aerosol components were retrieved and discussed in Sect.4.2
of the manuscript, $\sigma_{sp,X,PM_{25}}(RH)$ values at 550 nm of aerosol components X were
then converted to 550 nm using measured scattering Ångström exponent. Contributions
of aerosol components to visibility degradation were thus calculated as:
 $Contribution = \frac{\sigma_{sp,X,PM_{25}}(RH, S50 nm)}{\sigma_{e_X}(550 nm)}$

389 5. Other Figures



Figure S11. Simulated relationships between VSE_{PM1} and Rsca using Mie theory through varying volume geometric mean D_{gv} of lognormal size distributions from 100 to 700 nm under different standard deviation (σ_g) conditions.



Figure S12. Histogram of ambient relative humidity (RH) during the observation
 period.



414 Figure S13. Aerosol light scattering enhancement measurements (fRH) at 525 nm from
415 13th to 26th February with RH range of 60-90%.

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