

Supporting Information for

Synergetic effect of NH<sub>3</sub> and NO<sub>x</sub> on the  
production and optical absorption of secondary  
organic aerosol formation from toluene  
photooxidation

Shijie Liu <sup>a</sup>, Dandan Huang <sup>b</sup>, Yiqian Wang <sup>a</sup>, Si Zhang <sup>a</sup>, Can Wu <sup>a</sup>, Wei Du <sup>a</sup>, Gehui Wang <sup>a,c,\*</sup>

<sup>a</sup> *Key Lab of Geographic Information Science of the Ministry of Education, School of Geographic Sciences, East China Normal University, Shanghai 210062, China*

<sup>b</sup> *State Environmental Protection Key Laboratory of Formation and Prevention of the Urban Air Pollution Complex, Shanghai Academy of Environmental Sciences, Shanghai 200233, China*

<sup>c</sup> *Institute of Eco-Chongming, 3663 North Zhongshan Road, Shanghai 200062, China*

Corresponding author: Prof. Gehui Wang, e-mail: [ghwang@geo.ecnu.edu.cn](mailto:ghwang@geo.ecnu.edu.cn)

Positive matrix factorization (PMF) is a receptor model and multivariate factor analysis tool (Paatero and Tapper, 1994; Paatero, 1997). Recently, the PMF model was used to provide better separation of different organic components through high-resolution (HR) mass spectra data (Liu et al., 2014). This model was expressed as below:

$$x_{ij} = \sum_p g_{ip} f_{pj} + e_{ij}$$

where  $i$  and  $j$  refer to values of  $j$  species in  $i$  samples, respectively,  $p$  is the number of factors, and used a least-squares fitting process, minimizing a quality of fit parameter.

In our study, CU AMS PMF Execute Tool v 3.04A, which was developed by Ulbrich et al. (Ulbrich et al., 2009), was used for the PMF analysis. High-resolution ion fragments at  $m/z$  from 12-160 were used. We generated the organic data matrices and the corresponding error matrices from PIKA v 1.15D. Ions were classified and down-weighted according to the signal-to-noise ratios (SNR).  $0.2 < \text{SNR} < 2$  was classified as the weak ions and down-weighted by a factor of 2,  $\text{SNR} < 0.2$  was bad ions and removed from the analysis. Since  $\text{O}^+$ ,  $\text{HO}^+$ ,  $\text{H}_2\text{O}^+$  and  $\text{CO}^+$  are related proportionally only to  $\text{CO}_2^+$  in the fragmentation table, the error values for each of these  $m/z$  were multiplied to avoid excessive weighting of  $\text{CO}_2^+$ . The data were analyzed using the PMF2 algorithm (Paatero et al., 2002) with  $f_{\text{peak}}$  varying between -1 and 1.

A summary of the PMF results is presented in Fig. S1-S3. After an extensive evaluation of the mass spectral profiles and time series of different number of factors and the rotational forcing parameter,  $f_{\text{Peak}}$ , the 2-factor solution with  $f_{\text{Peak}} = 0$  was chosen for toluene SOA. The OA components of the 2-factor solution solved under different  $f_{\text{Peak}}$  values show very similar mass spectral patterns.

The direct comparisons of the mass spectra and time series of 3-factor solution are shown in Fig. S4. The 3-factor solution splits the High-nitrogen OA (Hi-NOA) into two components for which we cannot offer a physically meaningful interpretation. While the results of 2-factor solution are also used in the familiar chamber study (Chen et al., 2021; Chen et al., 2019). We therefore choose the 2-factor solution.

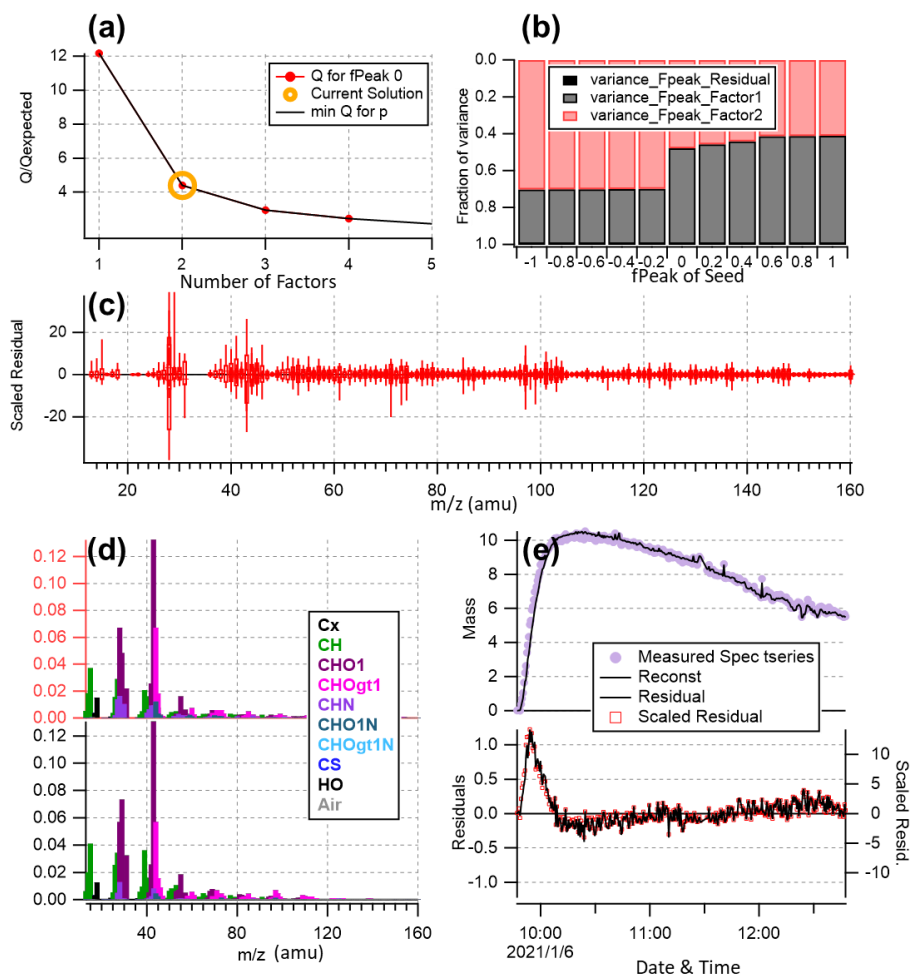


Fig. S1 The 2-factor solution for the toluene OH-oxidation in the presence of NH<sub>3</sub>.

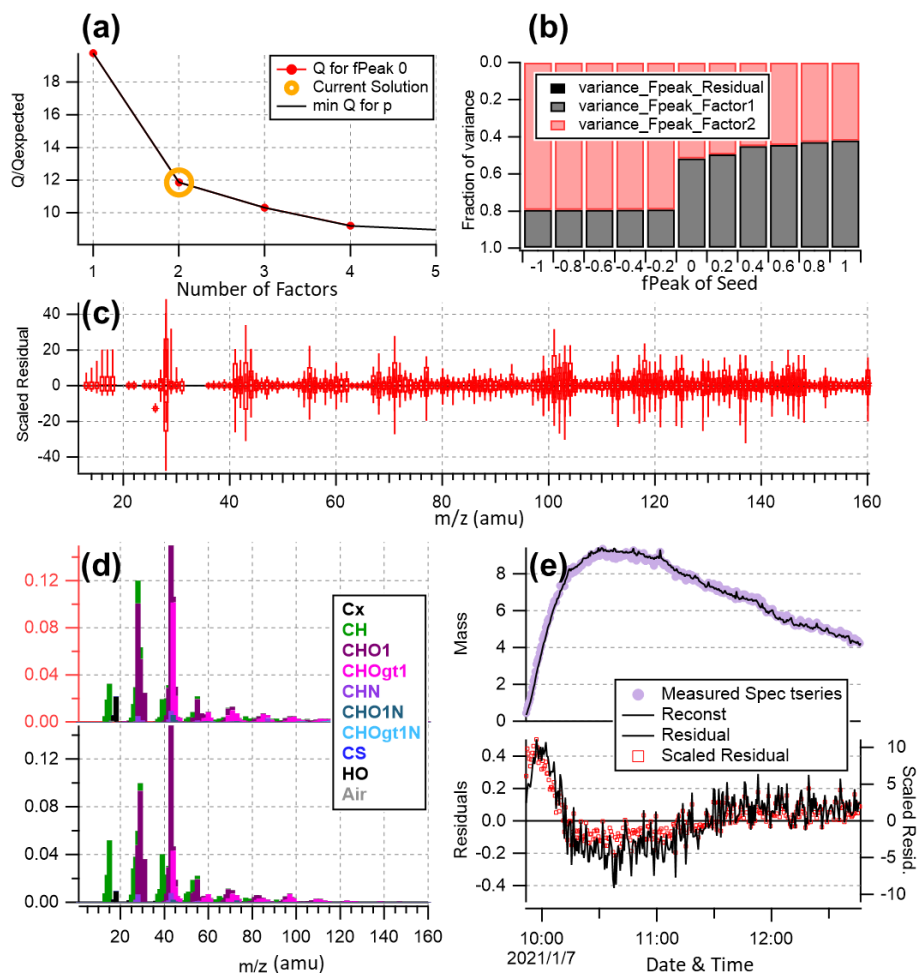


Fig. S2 The 2-factor solution for the toluene OH-oxidation in the presence of NO<sub>x</sub>.

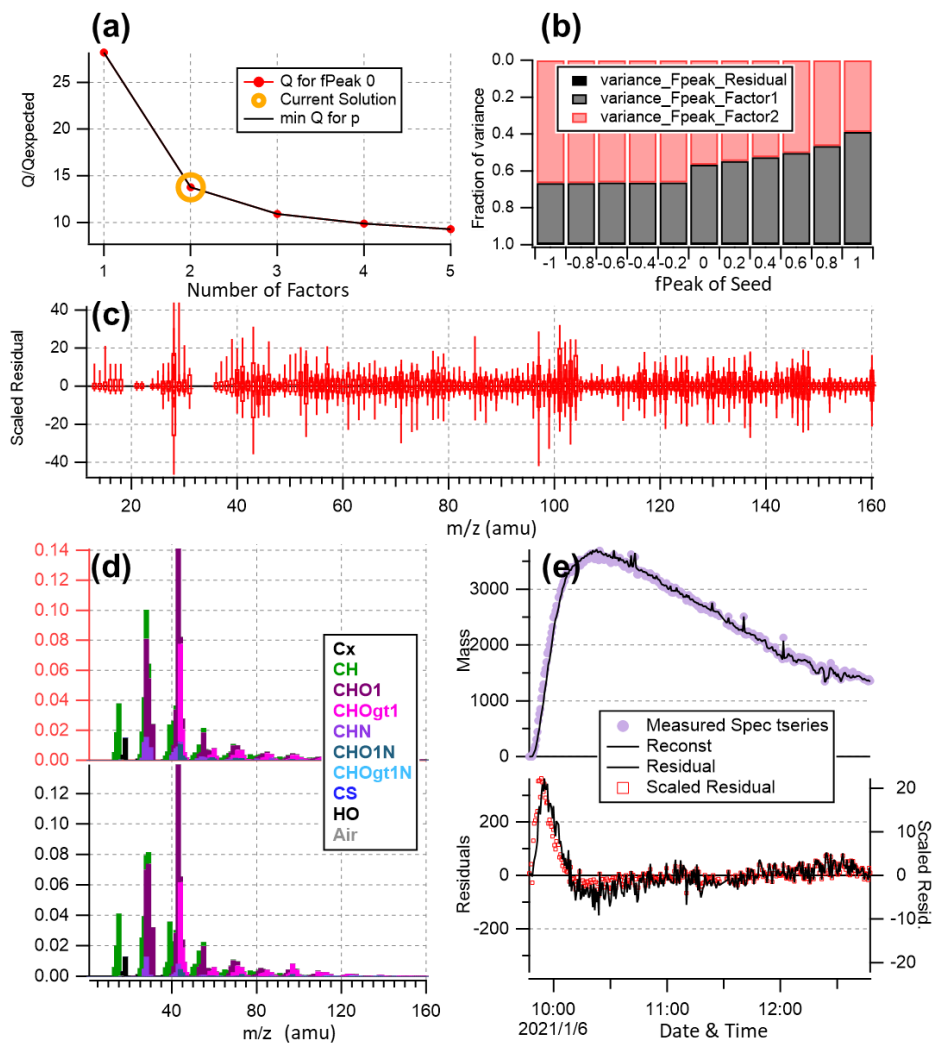


Fig. S3 The 2-factor solution for the toluene OH-oxidation in the presence of both NO<sub>x</sub> and NH<sub>3</sub>.

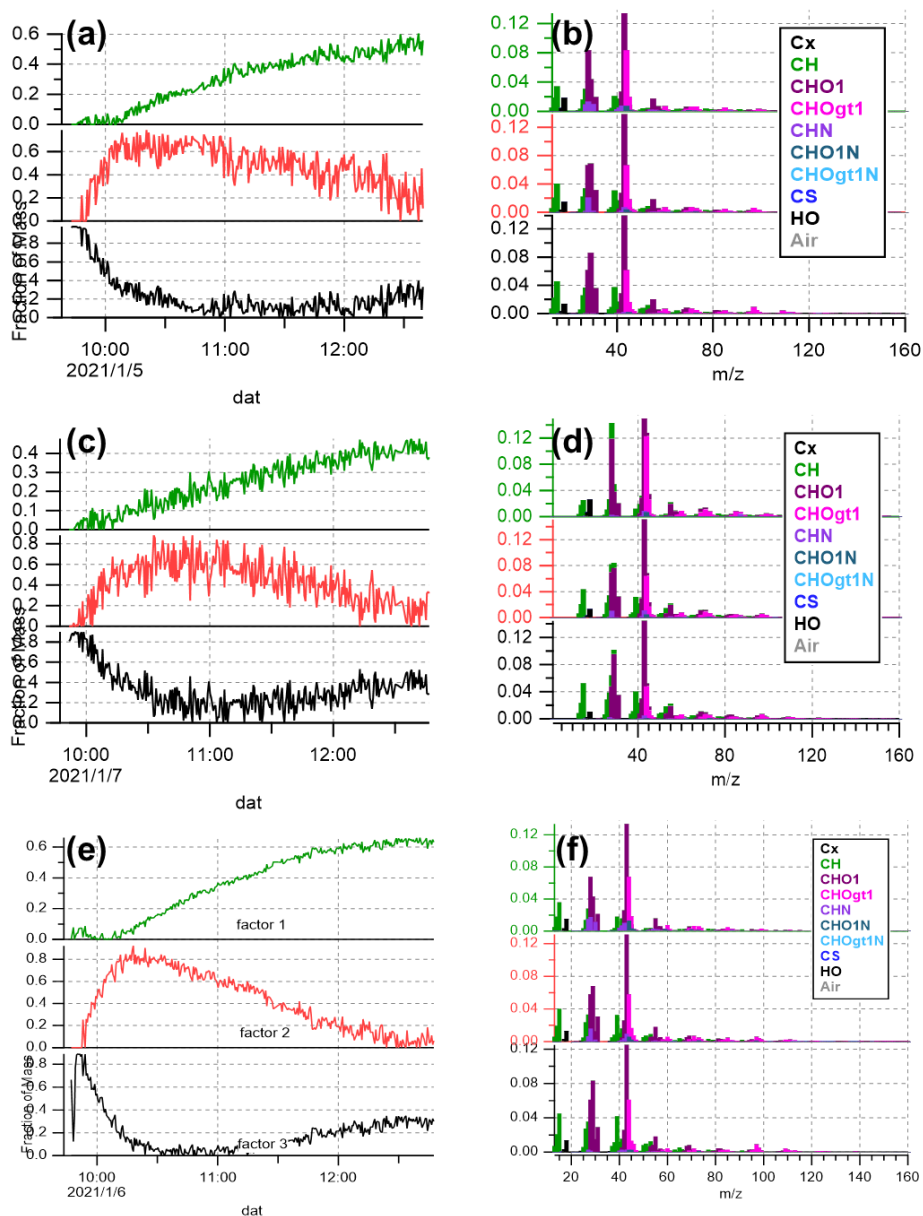


Fig. S4 (a), (c), and (e): High resolution mass spectra 3-factor solution for the Exp.2, 3, and 4, respectively. (b), (d), and (f): Time series of mass concentration of OA in each factor.

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