

Key Points:

- OC correlated well with ozone during a photochemical pollution episode
- The OC atmospheric evolution was explained by a parameterization method
- The photochemical consumption of VOCs accounted for 67% of SOC formation

Supporting Information:

- Supporting Information S1

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Estimation of Secondary Organic Aerosol Formation During a Photochemical Smog Episode in Shanghai, China

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Abstract Secondary organic aerosols (SOA) are formed through diverse processes in the atmosphere, among which photochemical processing is one important pathway. SOA formation was studied based on one heavy photochemical smog episode in summer in Shanghai. During the pollution episode, ozone and organic carbon (OC) increased simultaneously with a strong positive correlation, which was complete opposite to the volatile organic compounds (VOCs) pattern but similar to that of VOC photochemical consumption. The OC evolution was explained well by a parameterization method based on the observation of OC and VOCs, and secondary OC (SOC) formation was derived, being comparable with the result based on elemental carbon (EC) tracer method. About 67% of SOC could be explained by the photochemical consumption of VOCs (mainly aromatics, ~93%) during the episode. The contribution of VOCs to SOC formation was also estimated from the available VOC emissions inventories, which was comparable with that based on VOCs observations in ambient. Some differences of VOC species contribution to SOC were found between the ambient observation-based and the emission-based results, and the contribution of C9 aromatics was underestimated in the emission inventory. This suggests that bias of speciation might exist in the current VOC emissions inventories. The present study highlights the importance of VOC oxidation for SOC formation in summer in Shanghai. More insights are needed to improve the accuracy of VOCs speciated emissions inventories.

1. Introduction

The formation of secondary organic aerosols (SOA) has been of great concern due to its large impacts on changing cloud properties and air quality (Hoyle et al., 2009). Generally, SOA are produced through the homogeneous (Claeys et al., 2004) and heterogeneous (Jang et al., 2002) reactions of volatile organic compounds (VOCs) and intermediate/semi-VOCs (Chan, et al., 2009; Robinson et al., 2007; Guo et al., 2020) as well as the aging of primary aerosols (Jimenez et al., 2009) in the atmosphere.

Although great strides have been made in the past decades, the estimation of SOA in the atmosphere is still highly uncertain (An et al., 2019; Ehn et al., 2014; Kanakidou et al., 2005; Shrivastava et al., 2017; Volkamer et al., 2006). It remains difficult to quantify the contribution of precursors in the ambient environment. Several field-based methods have been developed to estimate the amount of SOA, including the tracer-yield method (Guo et al., 2012; Kleindienst et al., 2007), the nonprimary organic carbon (OC) method (the receptor model) (Heo et al., 2015; Yuan et al., 2006; Zheng et al., 2002), the non-biomass burning water-soluble organic carbon method (Weber et al., 2007) and the elemental carbon (EC)-tracer method (or the OC/EC ratio method) [Cao et al., 2007; Song Guo et al., 2014; Turpin & Huntzicker, 1995; Zhang et al., 2008]. Among these methods, the receptor model can be used to apportion SOA in submicron particulate matter (PM_{1.0}) at a high time resolution from aerosol mass spectrometry (AMS) data (Hu et al., 2013;

Huang et al., 2012; Huang et al., 2014; Li et al., 2015; Sun et al., 2013; Zhang, 2005), while the tracer-yield method provides the only way of identifying precursors of SOA but with a low time resolution due to the complexity of tracer measurements.

In fact, both SOA and primary organic aerosols (POA) evolve simultaneously with the gaseous precursors in the atmosphere (Palm et al., 2018). The formation and loss of SOA interact with the aging and loss of POA and the gaseous precursors. The processing of organic aerosols (OA) and the potential gaseous precursors have been investigated in many field studies around the world (Ait-Helal et al., 2014; de Gouw et al., 2008; Gao et al., 2019; Hayes et al., 2013; Hayes et al., 2015; Jimenez et al., 2009; Kleinman et al., 2016; Koss et al., 2015; Sjostedt et al., 2011; Slowik, 2010; Weber et al., 2007; Yuan et al., 2013; Zhou et al., 2014). These studies have shown the amount of SOA to be several times that of the initial POA concentrations after a few hours of photochemical aging and have found a corresponding decrease in the concentrations of the gaseous precursors. These findings demonstrate the impact of the photochemistry of precursors on the formation of SOA.

Shanghai, the largest megacity in China with strong emissions of air pollutants (Fu et al., 2013; Li et al., 2014; Li et al., 2017), has suffered from severe pollution of fine particulate matter ($PM_{2.5}$) driven to a large extent by secondary formation (Huang et al., 2014; Wang et al., 2016) and increasing ozone concentration in the atmosphere (Li et al., 2019). Previous studies revealed SOA formation was enhanced by increasing atmospheric oxidizing capacity in north of China (Feng et al., 2019; Xu et al., 2017). This suggests the complex interaction and formation mechanisms of secondary pollutants. SOA and ozone share one kind of important precursors, while the impact of photochemical processing on the formation of SOA in Shanghai is unclear and the contribution of VOCs to SOA formation is ambiguous.

This study focuses on a pollution episode of $PM_{2.5}$ that occurred simultaneously with an O_3 episode in Shanghai in the summer of 2013. The evolution of OC and photochemical loss of VOCs was studied, and accordingly secondary OC (SOC) formation was derived, which was further compared with the estimation of SOC by the elemental carbon (EC) tracer method. The contribution of VOC oxidation to SOC formation was studied based on both the VOCs measurements in ambient and their emission inventory. Finally, the implication of SOC formation in Shanghai was discussed.

2. Experimental Methods

All measurements were conducted on the rooftop of a five-story building at a height of about 15 m from the ground in the Shanghai Academy of Environmental Sciences (SAES, 31.17°N, 121.43°E) from 2 to 12 August 2013, which is located in the southwest of the central urban area of Shanghai, China. A detailed description of the monitoring site is available from our previous studies (Qiao et al., 2014; Wang et al., 2013; Wang et al., 2016).

The mass concentration of $PM_{2.5}$ was measured with a resolution of 1 hr by an online particulate monitor (FH 62 C14 series, Thermo Fisher Scientific Inc.) using beta attenuation equipped with a verified $PM_{2.5}$ cyclone. The hourly concentrations of the carbonaceous species in $PM_{2.5}$ were measured by a semicontinuous OC/EC analyzer (model RT-4, Sunset Laboratory Inc.) equipped with an upstream parallel-plate organic denuder. The detailed principle and operation of these instruments have been described in detail elsewhere (Wang et al., 2015; Wang et al., 2016).

Fifty-five VOC species were continuously measured every 30 min by two online high-performance gas chromatograph with flame ionization detector (GC-FID) systems (Chromato-sud airmoVOC C2-C6 #5250308 and airmoVOC C6-C12 #2260308, France). The detection limits for VOC species range from several tens to hundreds pptv, as listed in supporting information Table S1. A detailed description of the measurements can be found in our previous study (Liu et al., 2019; Wang et al., 2013).

The trace gases—nitrogen oxides (NO_x), carbon monoxide (CO), and ozone (O_3)—were routinely measured with a resolution of 1 min. All gas analyzers were calibrated using multipoint calibration. Meteorological parameters including wind speed, temperature, relative humidity, and solar radiation were also automeasured simultaneously during the pollution episode.

3. Results and Discussions

3.1. Overview of the Photochemical Pollution Episode

An episode of heavy photochemical smog occurred in Shanghai during the period of 2–12 August 2013, as shown in Figure S1. During this episode, the daily maximum of solar radiation reached 800 W/m^2 for a continuous period of over 7 days, and the daily peak temperatures were over 35°C for a continuous period of over 11 days. The 24 hr back trajectories during the period were shown in Figure S2. As indicated, the air masses during the period were mostly from the south to southwest of Shanghai, which meant there were no significant variation of the meteorology. Meanwhile, the back trajectories during 5 to 11 August (O_3 pollution episode with the daily maximum 1 h- O_3 concentration exceeding 100 ppbv) were shorter than those of other days, suggesting the air mass was more stagnant during 5 to 11 August. The measured wind speed was $4.0 \pm 1.5 \text{ m/s}$ on average during the whole period, lower than the mean of the wind speed (4.5 m/s) in summer (June, July, and August) (Wang et al., 2016), and especially the wind speed during the O_3 pollution episode was as low as $3.5 \pm 1.4 \text{ m/s}$. These meteorological conditions, combined with the relatively stagnant air mass and low humidity, hindered the dilution of air pollutants and induced the formation of photochemical smog. Consequently, for a continuous period between 5 and 11 August, the daily maximum 1 h- O_3 concentration exceeded Class II of China National Air Quality Standards (CNAQAS) for hourly average of $200 \mu\text{g/m}^3$ (~100 ppbv). The increases in VOC and NO_x concentrations played an important role in the high concentration of ozone during 5–11 August. The concentration of VOCs was as high as ~60 ppbv, doubling the annual average of ~27 ppbv (Wang et al., 2013). The diurnal patterns of VOCs and NO_x were completely opposite to that of O_3 , demonstrating chemical losses of precursors contribute to the formation of O_3 , as shown in Figures 1 and 2.

In the meantime, the $\text{PM}_{2.5}$ mass concentration reached high levels during the episode, averaging $48 \mu\text{g/m}^3$ with hourly maximum of $94 \mu\text{g/m}^3$. These values are significantly higher than those before and after the episode. The mass concentration of $\text{PM}_{2.5}$ was higher in daytime than nighttime, as shown in Figure 1, and showed a good agreement with O_3 , with a correlation coefficient (R) of 0.65, as shown in Figure 3. According to a study in the western Yangtze River Delta region, these two pollutants are negatively correlated under low temperatures but positively correlated under high temperatures (Ding et al., 2013). This suggests the importance of the photochemical formation of $\text{PM}_{2.5}$ during a pollution episode.

OC accounted for 28.9% of $\text{PM}_{2.5}$, making it the most abundant component during the episode. If an organic matter OM/OC ratio of 1.8 is applied to estimate the mass of OA, the mass fraction of OA in $\text{PM}_{2.5}$ would be as high as 42%, which is comparable to the 48% reported in a previous study (Huang et al., 2014). There was a significant peak of the diurnal pattern of the OC/EC ratio during noontime, as shown in Figure 1. The correlation between OC and O_3 was even stronger than that between O_3 and $\text{PM}_{2.5}$, and the correlation coefficient (R) was 0.67, as shown in Figure 3. This suggests that the photochemical formation played more important roles in OC than in $\text{PM}_{2.5}$ during the episode.

3.2. SOC Formation From VOCs Oxidation (SOC_{voc})

3.2.1. Photochemical Consumption of VOCs

As mentioned in section 3.1, the diurnal pattern of VOCs was completely opposite to that of O_3 , and the concentration during nighttime was 18% higher than that of daytime (Table S1), suggesting the strong chemical loss of VOCs inducing the formation of photochemical pollution, as shown in Figure 2. VOC photochemical consumption was calculated in this section.

The photochemical consumption of VOCs can be described by equation 1 below, with the assumption that the oxidation of OH radicals dominates the loss of VOCs in the ambient environment. Generally, the photochemical age of the air masses could be described by the measured ratio of two VOC species with the similar emission sources but with different atmospheric reactivity and their measured uncertainties should relatively be low. Several ratios have been used to estimate the photochemical age, such as ethylbenzene to m,p-xylene, benzene to toluene, benzene to ethylbenzene, benzene to toluene, etc (de Gouw, 2005; Wang et al., 2013; Yuan et al., 2013). Correlations of benzene to toluene, benzene to ethylbenzene, and benzene to toluene with ozone were relatively bad, suggesting the sources of each couple of VOC species were not very similar in Shanghai. The photochemical age (t) of the air mass was estimated from the measured concentration ratios of ethylbenzene (E) to m,p-xylene (X) in this study, and the diurnal pattern of E/X reflected

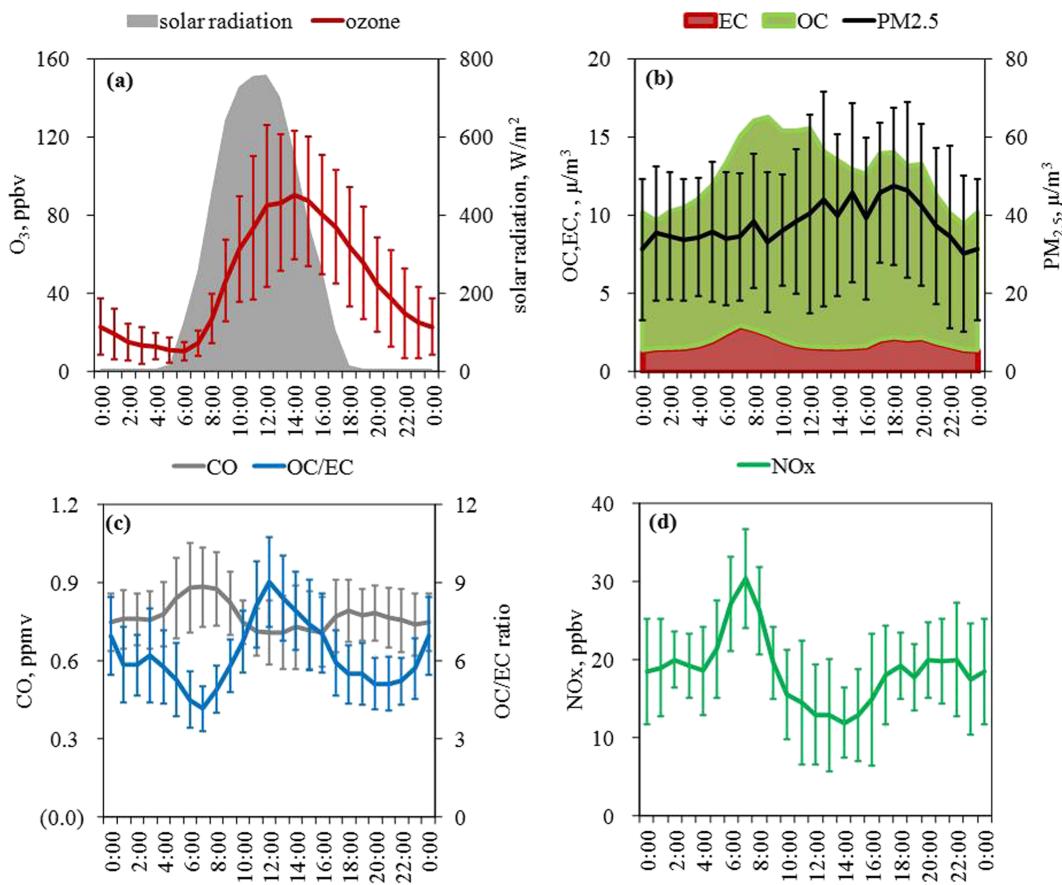


Figure 1. Diurnal variations of (a) ozone concentration and solar radiation, (b) PM_{2.5}, OC, and EC concentration, (c) CO concentration and the ratio of OC/EC, and (d) NO_x concentration of the whole campaign.

the evolution of photochemistry in the present study as shown in Figure 2a, which has been proved to be suitable in Shanghai in our previous studies (Gao et al., 2019; Wang et al., 2013) according to equation 2 below.

$$VOC_{i,consumed} = VOC_{i,t} \times (\exp(k_i[OH]\Delta t) - 1) \quad (1)$$

$$\Delta t = \frac{1}{[OH](k_E - k_X)} \times \left(\ln \left\{ \frac{[E]}{[X]} \right\}_{t=t_0} - \ln \left\{ \frac{[E]}{[X]} \right\}_{t=T} \right) \quad (2)$$

Here, $VOC_{i,consumed}$ and $VOC_{i,t}$ are the consumption and measured concentrations of VOC_i, respectively. k_i is the reaction rate constants of VOC_i with OH radicals. k_E (7.0×10^{-12} cm³ molecule⁻¹ s⁻¹) and k_X (18.7×10^{-12} cm³ molecule⁻¹ s⁻¹) are the reaction rate constants of ethylbenzene (E) and m,p-xylenes (X) with OH radicals (Atkinson et al., 2006). [OH] is the concentration of OH radicals, which is not necessary for the calculation of $VOC_{i,consumed}$. $([E]/[X])_{t=t_0}$ is the ratio between the initial mixing ratios of ethylbenzene and m,p-xylenes in the fresh emissions. Generally, there were several methods to get $([E]/[X])_{t=t_0}$. (1) The first one is using the minimum of the measured ratios of $[E]/[X]$ in ambient but with enough data. (2) The second one is using the emission ratio of $[E]/[X]$ in the local sources. (3) The third one is using the average measured ratio of $[E]/[X]$ during the period with negligible photochemical processing, like 0:00–4:00. In the present study, considering the studied period was only 11 days, the $([E]/[X])_{t=t_0}$ was using the lowest 10% value (0.6) of measured ratios of $[E]/[X]$ during summer (July, August, and September) in 2013, as shown in Figure S3. $([E]/[X])_{t=t}$ is the ratio of $[E]$ to $[X]$ at time t .

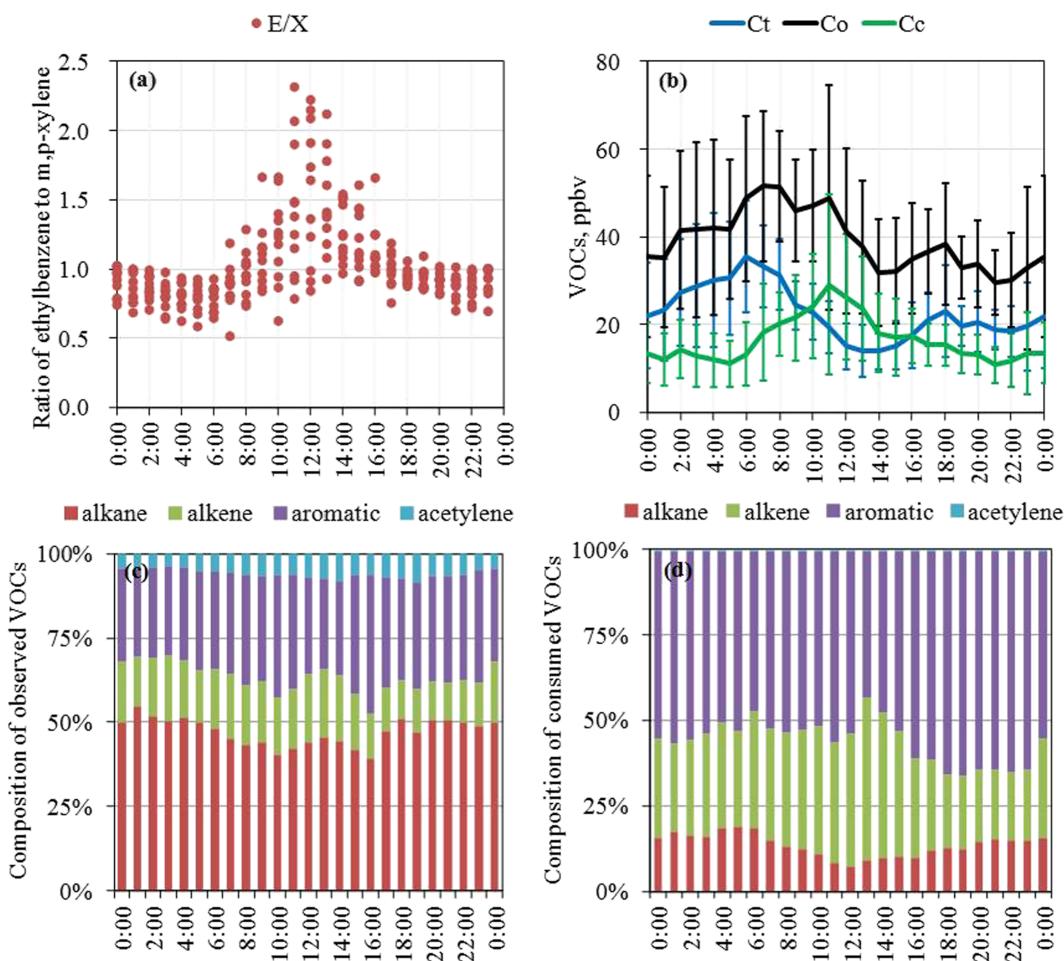


Figure 2. Diurnal pattern of (a) the ratio of ethylbenzene to m,p-xylene, (b) observed VOC concentrations (Ct) and chemical consumption concentrations (Cc), and chemical compositions of (c) measured and (d) consumed VOCs of the whole campaign.

Accordingly, the diurnal pattern of the concentration of VOC photochemical consumption (Cc) was shown in Figure 2b, with a peak at 11:00 with an average of 29.1 ± 20.1 ppbv, which was earlier than the time (at 14:00) with the lowest measured concentration (Ct). Based on equation 1, the photochemical loss of a given VOC was calculated using its observed concentration (Ct) and OH exposure ($[\text{OH}]\Delta t$) which was dependent

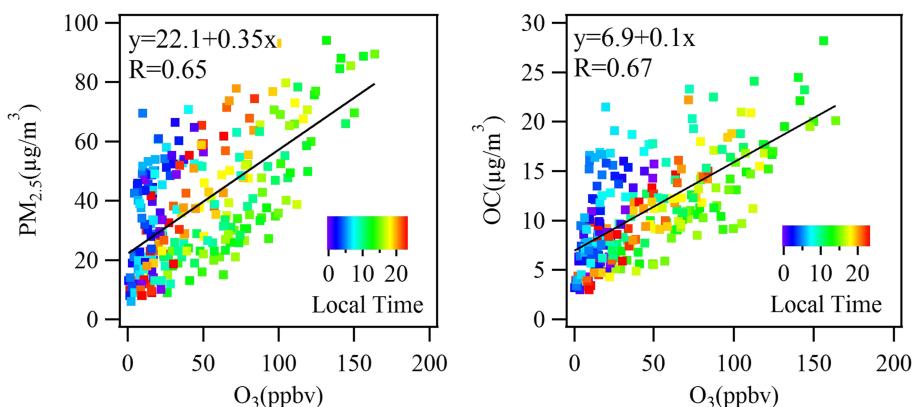


Figure 3. Scatterplots of (left) PM_{2.5} and (right) OC with O₃ concentrations color-coded by local time.

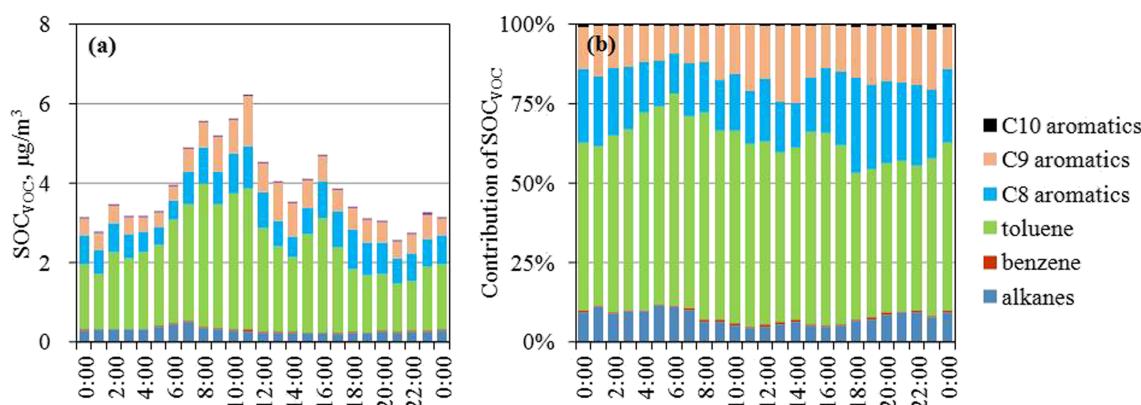


Figure 4. The average diurnal patterns of (a) SOC_{VOC} and (b) its compositions.

on the observed ratio of ethylbenzene to m,p-xylene ($([E]/[X])_{t=t}$). It should be pointed out that there were some variations of the results, as shown in Figure 2b, which was probably due to the limited data in this study. On average, C_c was 17.2 ± 8.0 ppbv, slightly lower than C_t (22.7 ± 8.5 ppbv). For the initial concentration (C_0), its diurnal pattern had a peak in the early morning, which probably resulted from the traffic and human activities as well as the low mixing layer boundary. The lowest concentration of C_0 happened around 14:00, mainly due to the favorable diffusion conditions.

Meanwhile, the species compositions of consumed VOCs were highly different from those of measured VOCs, as shown in Figures 2c and 2d. Aromatics dominated the consumed VOCs, accounting for 55%, followed by alkene (32%) and alkane (12%). This means that aromatics and alkenes were more important precursors of photochemical processing due to their high reactivity than alkanes which dominated the measured VOCs.

3.2.2. SOC_{VOC} Estimation

VOC photochemical consumption plays an important role in SOC formation in urban areas (eg., de Gouw et al., 2008; Ding et al., 2012; Emanuelsson et al., 2013; Gao et al., 2019; Guo et al., 2013; Spracklen et al., 2011). Based on the VOC photochemical consumption, the corresponding formation of SOC (SOC_{VOC}) was calculated using equation 3 as follows:

$$SOC_{VOC} = SOA/1.8 = \left(\sum_i VOC_{i, \text{consumed}} \times Y_i \right) / 1.8 \quad (3)$$

where $VOC_{i, \text{consumed}}$ is the consumption of VOC_i as mentioned above; Y_i is the SOA yield of VOC_i oxidation which is determined from chamber studies; and the ratio of SOC to SOA is assumed to be 1.8.

The SOA yields have been published for most VOCs, and are 0 for small alkanes and alkenes, low for the higher (>C8) and cyclic alkanes, and the highest for aromatics (Seinfeld & Pandis, 2006). Previous studies showed that SOA yields of hydrocarbons heavily depended on NO_x levels, mainly due to the competition reactions of RO₂ radical with NO and HO₂ radical (Ng et al., 2007). The most recent studies found that the dimerization of RO₂ radicals in the presence of autoxidation pathway at low NO_x condition also play important role in the higher SOA yields (Praske et al., 2018; Schwantes et al., 2017). SOA yields of most aromatics under low-NO_x conditions were significantly higher than those under high-NO_x conditions (Ng et al., 2007). The average NO_x concentration in the ambient was 2.9 ± 2.2 ppbv in this study, under which the chemistry occurring was consistent with that under high NO_x conditions created in chamber studies. Chamber studies showed that SOA yields of aromatics under high-NO_x conditions could be described using an empirical relationship based on gas-particle partitioning of two semivolatile products (Odum et al., 1996). The SOA yield from a given VOC was then used in the parameterization method and determined from ambient temperature ($T = 306.5$ K in this study) and organic aerosol mass loading (OM = $19.0 \mu\text{g m}^{-3}$ in this study). The vapor wall loss has been recognized to be underestimated largely, which resulted in the underestimation of chamber yields in previous studies. The vapor wall loss correction factors of SOA yields were

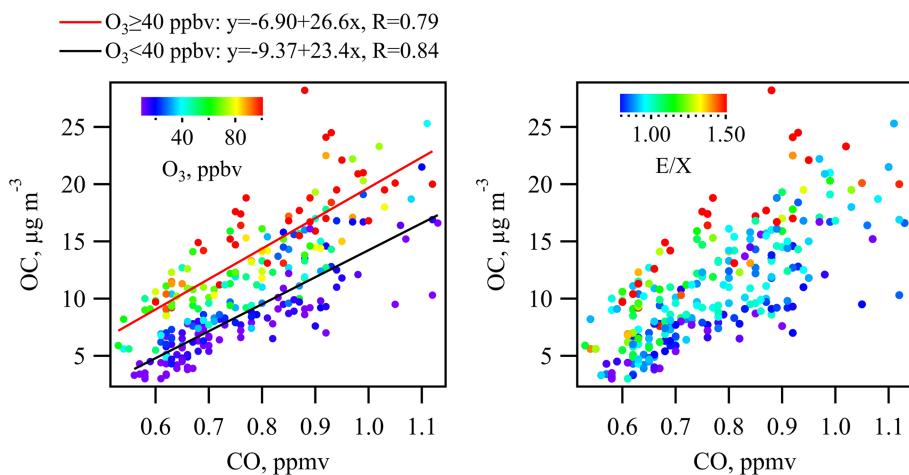


Figure 5. Scatterplots of OC and CO color-coded by O₃ (left) and the E/X ratio (right).

obtained from Zhang et al. (2014) in this study. Due to the lack of parameters in the two-product relationship and NO_x dependence information for alkanes and cycloalkanes in the literature, SOA yields of these compounds are directly adapted from Lim and Ziemann (2009) and further corrected for vapor wall losses (Zhang et al., 2014). Sensitivity analysis of the selections of temperature and OC mass concentration to the yields calculation was tested using the hourly observed data and the impacts were found to be negligible (as shown in Figure S4). The calculated SOA yield of each VOC species were tabulated in Table S2.

SOC_{VOC} determined from VOC oxidations was $3.9 \pm 1.0 \mu\text{g/m}^3$ on average during the whole period. The oxidation of aromatics dominated SOC_{VOC} formation, accounting for 93% SOC, as shown in Figure 4. Specifically, toluene was the most important VOC precursor of SOC_{VOC}, accounting for 57% of SOC_{VOC}, followed by C8 (19%) and C9 aromatics (16%). SOC_{VOC} took up 35% of OC in PM_{2.5} during the present episode. But how much did VOCs contribute to SOC during summer in Shanghai? To answer this question, SOC formed during the episode was estimated as described below.

3.3. SOC Estimation Via a Parameterization Method (SOC)

3.3.1. SOC Estimation

Figure 5 showed scatterplots of OC versus CO color-coded by O₃ and the E/X ratio. As shown, OC correlated with CO well, and the correlation coefficients (R) were 0.84 (O₃ concentration <40 ppbv) and 0.79 (O₃ concentration ≥ 40 ppbv), respectively. The average ratio of OC to CO (~27 µg m⁻³ ppmv⁻¹ for the data with O₃ concentration higher than 40 ppbv) was high during the periods with high concentrations of O₃ and high ratios of E/X indicating strong photochemistry, whereas this ratio (~23 µg m⁻³ ppmv⁻¹ for the data with O₃ concentration lower than 40 ppbv) was low in the air mass with relatively low O₃ levels and low ratios of E/X. These observations further confirmed the large contribution of photochemical formation to OC.

A semiempirical relationship between OC/ΔCO and the photochemical age, Δt, of urban plumes has been derived, which can explain a significant portion of the variation in the OC data (de Gouw, 2005; de Gouw et al., 2008; Yuan et al., 2013), where ΔCO was CO subtracted the background concentration to exclude the influences of emitted and transported OC. Accordingly, the measured concentrations of OC can be separated into primary emission, secondary formation and background concentration using equation 4 below:

$$\begin{aligned}
 OC &= ER_{OC} \times ([CO] - 0.6) \times \frac{\exp(L_{OC}\Delta t)}{\exp(-k_{OC}[\text{OH}]\Delta t)} \\
 &+ ER_{precursor} \times Y_{OC} \times ([CO] - 0.6) \times \frac{P_{OC}}{L_{OC} - P_{OC}} \times \frac{\exp(-P_{OC}\Delta t) - \exp(-L_{OC}\Delta t)}{\exp(-k_{CO}[\text{OH}]\Delta t)} \\
 &+ [bg]
 \end{aligned} \tag{4}$$

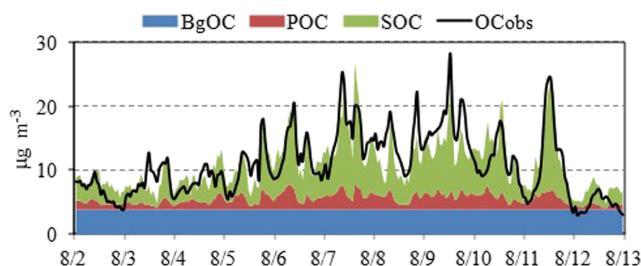


Figure 6. Time series of measured and simulated concentrations of OC during the episode. BgOC, POC, and SOC were the background concentration of OC, primary OC, and secondary OC; OC_{obs} was the measured OC.

the summer of 2016 in Shanghai by Tan et al. (2019) from the observation-based model. Accordingly, the first item in equation 4 represents the primary emission of OC and its removal (denoted by POC). The second item represents the secondary formation of OC from the available precursors and its removal (denoted by SOC). The unknown parameters are ER_{OC}, ER_{precursor} × Y_{OC}, L_{OC}, P_{OC}, and [bg].

In order to obtain more stable and reasonable fits, the initial settings of these five unknown parameters were given empirically. The lifetime of OC in the atmosphere fell in the range of 3–10 days (Koch, 2001; Millet et al., 2004), and correspondingly the value of L_{OC} varied from 1.2×10^{-6} to 3.9×10^{-6} s⁻¹. The initial setting of production rate of OC (P_{OC}) was similar to that of L_{OC}. For the emission ratio of OC to CO (ER_{OC}), the result from the emission inventory of Shanghai was $\sim 8 \mu\text{g m}^{-3}$ ppmv⁻¹ (Fu et al., 2013), and the average measured ratio in the present study was $13.9 \mu\text{g m}^{-3}$ ppmv⁻¹, and accordingly the initial setting of ER_{OC} was $5 \mu\text{g m}^{-3}$ ppmv⁻¹. The initial settings of the background concentration of OC (BgOC) and the product (ER_{precursor} × Y_{OC}) of emission ratio of SOC precursor (ER_{precursor}) and its yield of SOC (Y_{OC}) referred to the results of Yuan et al. (2013) study.

After setting the initial values, the sensitivity of each parameter was tested to obtain reasonable fits. As shown in Figure S7, as OC lifetime varied from 4 days to 10 days, the fitted results of the other parameters were quite stable. To present the average results, the lifetime of OC was set as 6 days, and thus the uncertainty of OC loss rate (L_{OC}) was estimated as the standard variation with the OC lifetime ranging from 1 to 10 days. The uncertainties of other parameters were the standard variations of their variations with the change of OC lifetime. All the fitting results and uncertainties were summarized in Table S3.

The background concentration of OC was fitted to be $3.8 \pm 0.4 \mu\text{g m}^{-3}$, which was higher than that of $2.4 \mu\text{g m}^{-3}$ on Changdao Island (Yuan et al., 2013). This was expected because the strong emissions around Shanghai city would have increased the background concentration of OC in Shanghai, whereas the emissions around Changdao Island would have been less strong seeing as it is located away from the city clusters. ER_{OC} was determined to be $5.8 \pm 0.5 \mu\text{g m}^{-3}$ ppmv⁻¹, which was lower than the emission result (Fu et al., 2013). ER_{precursor} × Y_{OC} and P_{OC} strongly coupled in the fit, and therefore hard to determine independently from each other.

The OC, POC, SOC, and BgOC were simulated based on the fitting parameters, as shown in Figure 6. On average, the correlation efficient (R^2) between measured (OC_{obs}) and simulated (OC_{sim}) concentrations was 0.723, as shown in Figure S9, and the average ratio between OC_{sim} and OC_{obs} was 1.1 ± 0.3 . Accordingly, the fractions of POC and SOC in total OC were 12% and 49%, respectively.

3.3.2. OC Evolution Along With the Photochemical Age

According to the fitting results, the evolution of the simulated OC with the photochemical age was presented in Figure 7. Photochemical age was estimated by equation 2 above. [OH] is the average OH radical concentration (1.7×10^6 molecule cm⁻³) estimated from the empirical equation as mentioned above. As shown in Figure 7c, the primary emissions and the background concentrations were a significant fraction of OC in the first few hours. The secondary formation continued for about 3 days, after which OC was almost exclusively from secondary sources. Considering the estimated photochemical ages during the whole period in this study were lower than 20 hr, we fitted the data with photochemical age lower than 20 hr in Figure 7c by

ER_{OC} and ER_{precursor} are emission ratios of OC and OC precursors to CO. L_{OC} and P_{OC} are the loss and formation rates of OC. Y_{OC} is the yield of precursors that form OC. [CO] is the concentration of CO, with the background concentration of 0.6 ppmv (Figure S5), which is determined from the intercept in the scatterplot of CO with benzene by the method of DeCarlo et al. (2010). k_{co} (0.24×10^{-12} cm³ molecule⁻¹ s⁻¹) is the reaction rate constant of CO with OH radicals (Atkinson et al., 2006). [bg] is the background concentration of OC (denoted by BgOC). The other parameters are the same as those in equation 1. [OH] is the average OH radical concentration (1.7×10^6 molecule cm⁻³) estimated from the empirical equation (Ehhalt & Rohrer, 2000). As shown in Figure S6, the average of the daily maximum concentration was about 6×10^6 molecule cm⁻³, comparable with the only published result (7×10^6 molecule cm⁻³) in the summer of 2016 in Shanghai by Tan et al. (2019) from the observation-based model. Accordingly, the first item in equation 4 represents the primary emission of OC and its removal (denoted by POC). The second item represents the secondary formation of OC from the available precursors and its removal (denoted by SOC). The unknown parameters are ER_{OC}, ER_{precursor} × Y_{OC}, L_{OC}, P_{OC}, and [bg].

In order to obtain more stable and reasonable fits, the initial settings of these five unknown parameters were given empirically. The lifetime of OC in the atmosphere fell in the range of 3–10 days (Koch, 2001; Millet et al., 2004), and correspondingly the value of L_{OC} varied from 1.2×10^{-6} to 3.9×10^{-6} s⁻¹. The initial setting of production rate of OC (P_{OC}) was similar to that of L_{OC}. For the emission ratio of OC to CO (ER_{OC}), the result from the emission inventory of Shanghai was $\sim 8 \mu\text{g m}^{-3}$ ppmv⁻¹ (Fu et al., 2013), and the average measured ratio in the present study was $13.9 \mu\text{g m}^{-3}$ ppmv⁻¹, and accordingly the initial setting of ER_{OC} was $5 \mu\text{g m}^{-3}$ ppmv⁻¹. The initial settings of the background concentration of OC (BgOC) and the product (ER_{precursor} × Y_{OC}) of emission ratio of SOC precursor (ER_{precursor}) and its yield of SOC (Y_{OC}) referred to the results of Yuan et al. (2013) study.

After setting the initial values, the sensitivity of each parameter was tested to obtain reasonable fits. As shown in Figure S7, as OC lifetime varied from 4 days to 10 days, the fitted results of the other parameters were quite stable. To present the average results, the lifetime of OC was set as 6 days, and thus the uncertainty of OC loss rate (L_{OC}) was estimated as the standard variation with the OC lifetime ranging from 1 to 10 days. The uncertainties of other parameters were the standard variations of their variations with the change of OC lifetime. All the fitting results and uncertainties were summarized in Table S3.

The background concentration of OC was fitted to be $3.8 \pm 0.4 \mu\text{g m}^{-3}$, which was higher than that of $2.4 \mu\text{g m}^{-3}$ on Changdao Island (Yuan et al., 2013). This was expected because the strong emissions around Shanghai city would have increased the background concentration of OC in Shanghai, whereas the emissions around Changdao Island would have been less strong seeing as it is located away from the city clusters. ER_{OC} was determined to be $5.8 \pm 0.5 \mu\text{g m}^{-3}$ ppmv⁻¹, which was lower than the emission result (Fu et al., 2013). ER_{precursor} × Y_{OC} and P_{OC} strongly coupled in the fit, and therefore hard to determine independently from each other.

The OC, POC, SOC, and BgOC were simulated based on the fitting parameters, as shown in Figure 6. On average, the correlation efficient (R^2) between measured (OC_{obs}) and simulated (OC_{sim}) concentrations was 0.723, as shown in Figure S9, and the average ratio between OC_{sim} and OC_{obs} was 1.1 ± 0.3 . Accordingly, the fractions of POC and SOC in total OC were 12% and 49%, respectively.

3.3.2. OC Evolution Along With the Photochemical Age

According to the fitting results, the evolution of the simulated OC with the photochemical age was presented in Figure 7. Photochemical age was estimated by equation 2 above. [OH] is the average OH radical concentration (1.7×10^6 molecule cm⁻³) estimated from the empirical equation as mentioned above. As shown in Figure 7c, the primary emissions and the background concentrations were a significant fraction of OC in the first few hours. The secondary formation continued for about 3 days, after which OC was almost exclusively from secondary sources. Considering the estimated photochemical ages during the whole period in this study were lower than 20 hr, we fitted the data with photochemical age lower than 20 hr in Figure 7c by

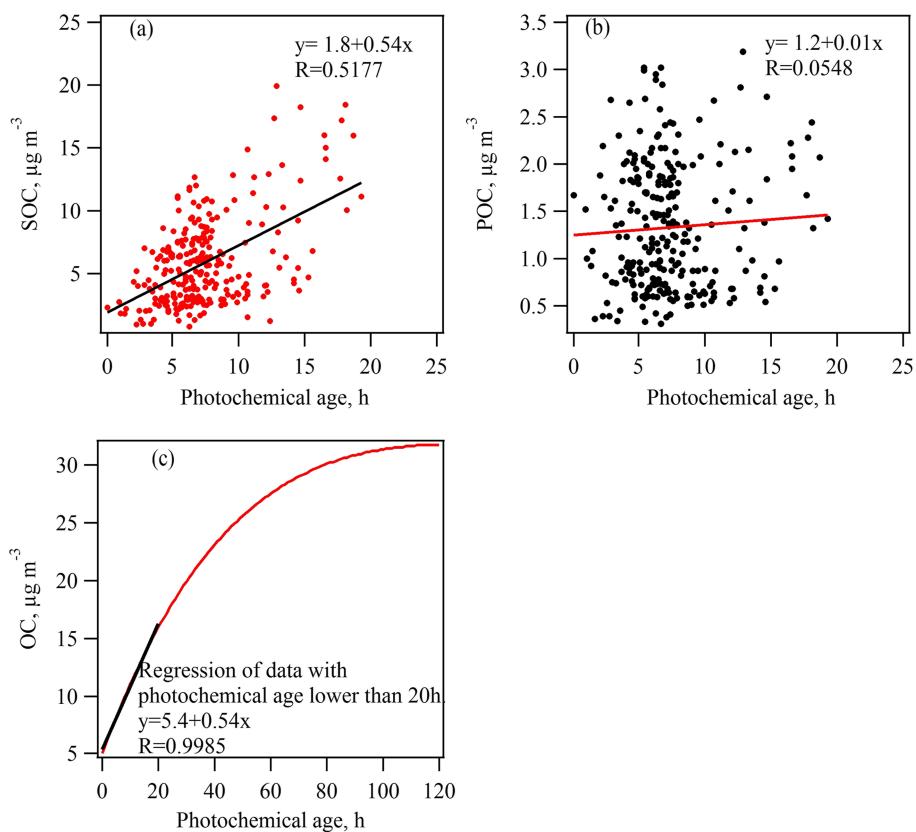


Figure 7. Scatterplots of the simulated (a) SOC and (b) POC with the photochemical age, and (c) the evolution of OC along with the photochemical age.

linear regression. The regression slope of OC versus photochemical age was $0.54 \mu\text{g m}^{-3} \text{ hr}^{-1}$, which was comparable with the regression slope of the simulated SOC versus photochemical age in this study (as shown in Figure 7a). The simulated SOC increased as the photochemical age increased, and presented a good correlation with the photochemical age with the correlation coefficient (R) of 0.52. As expected, bad correlation of POC with the photochemical age was obtained.

3.4. SOC Estimation by EC Tracer Method (SOC')

Elemental carbon (EC) has been widely used as a tracer to track the portion of co-emitted primary OC and, by extension, to estimate SOC from ambient observations of EC and OC (Turpin & Huntzicker, 1995). Key to this EC tracer method is to determine an appropriate OC/EC ratio that represents primary combustion emission sources (i.e., $(\text{OC/EC})_{\text{pri}}$) at the observation site. The approaches include regressing OC against EC within a fixed percentile of the lowest OC/EC ratio data (usually 5–20%) or relying on a subset of samples with low photochemical activity and dominated by local emissions. Considering the limited measurement of the studied period, the present study estimated $(\text{OC/EC})_{\text{pri}}$ (2.5) by the regression of OC and EC with the lowest 10% OC/EC ratios in August and September of 2013, as shown in Figure S10. Accordingly, SOC' was estimated as $6.2 \pm 3.4 \mu\text{g/m}^3$.

3.5. Comparisons of SOC Calculated From Different Methods

Figure 8 compares the SOC calculated from the three methods mentioned above. As shown, SOC was $5.8 \pm 3.6 \mu\text{g/m}^3$ derived from the parameterization method in section 3.3 and $6.2 \pm 3.4 \mu\text{g/m}^3$ derived from the EC tracer method in section 3.4. These two values are comparable. SOC_{voc} ($3.9 \pm 1.0 \mu\text{g/m}^3$) determined from the yield method in section 3.2 could explain 67.2% of total SOC, which was much higher than that (29%) of clean period in Shanghai in summer (Gao et al., 2019). It suggested that the photochemical oxidation of VOCs, especially aromatics, by OH radicals played important roles in SOC formation during the photochemical pollution episode in summer in Shanghai.

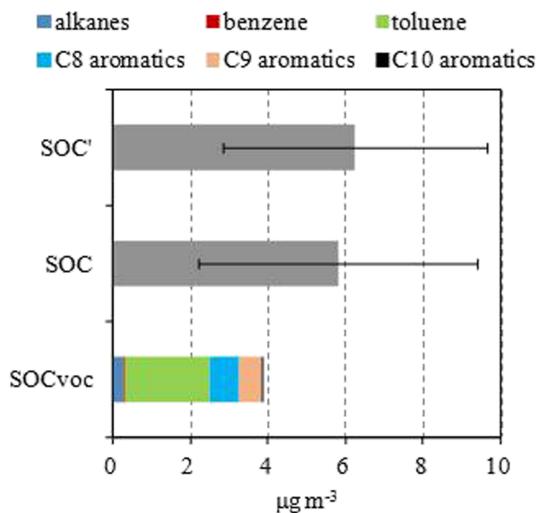


Figure 8. Comparisons of SOC calculated by different methods (the parameterization method for SOC; the EC tracer method for SOC'; the yield method for SOCvoc).

As shown, toluene SOA dominated SOCvoc. One important reason was due to the high SOA yield of toluene in the calculation. We corrected the yield of toluene SOA by a factor of 10 owing to vapor wall losses, but applied a wall-loss correction factor in the range of 1.16–1.25 for all the other alkanes and aromatic compounds (Gao et al., 2019; Zhang et al., 2014). There has been some evidence showing that toluene photochemistry that leads to SOA formation is unique. For example, Ng et al. (2007) reported that toluene yield is twice as much as xylenes. While benzene SOA yields are highest among the three, the degradation rate of benzene in the atmosphere is an order of magnitude slower than that of toluene so its contribution to SOA formation is considered much less important. Liu et al. (2016) measured the evaporation rates of particulate constituents of toluene SOA under all humidity ranges (0–50%) are a factor of 10 slower than those of SOA materials produced from xylenes.

3.6. Comparisons With the Previous Studies

For comparison with the previous studies, the OC/ΔCO ratio is used to evaluate the contribution of SOC formation, where ΔCO is CO subtracted the background concentration (0.6 ppmv in this study, as mentioned in section 3.3) to exclude the influences of emitted and transported OC

(DeCarlo et al., 2010). With the formation of SOC, the OC/ΔCO ratio increased and thus the enhancement of the OC/ΔCO ratio, in other words Δ (OC/ΔCO), could represent the formation of SOC. Based on the diurnal pattern of OC/ΔCO ratio, the ratio rose from a minimum of $30.6 \mu\text{g m}^{-3} \text{ ppmv}^{-1}$ at 6:00 to a maximum of $59.7 \mu\text{g m}^{-3} \text{ ppmv}^{-1}$ at 12:00 on average in this study, with an increase of $28.5 \mu\text{g m}^{-3} \text{ ppmv}^{-1}$. The Δ (OC/ΔCO) estimated in this study was higher than those of Beijing urban ($7.8 \mu\text{g m}^{-3} \text{ ppmv}^{-1}$) [W Hu et al., 2017], Guangzhou rural ($20 \mu\text{g m}^{-3} \text{ ppmv}^{-1}$) [W W Hu et al., 2012], and Changdao rural ($10.4 \mu\text{g m}^{-3} \text{ ppmv}^{-1}$) (Yuan et al., 2013) reported previously, as presented in Figure 9. It suggested, again, the strong photochemistry played very important roles in SOC formation during the ozone pollution episode in Shanghai.

The enhancement of OC/ΔCO ratio obtained above was higher than the average ratio of SOC/ΔCO ($22.2 \pm 5.7 \mu\text{g m}^{-3} \text{ ppmv}^{-1}$) of the whole period, as shown in Figure 9. The average ratio of SOC_{voc}/ΔCO was $14.9 \pm 5.0 \mu\text{g m}^{-3} \text{ ppmv}^{-1}$, accounting for 66.9% of SOC/ΔCO which was close to the proportion of SOC_{voc} in SOC mentioned in section 3.5. For comparison, the ratio of SOC_{voc}/CO was also estimated from the emissions inventory of VOCs and CO in Shanghai. The CO emissions inventory in 2013 was used as the average of data from the MEIC in 2012 and 2014 developed by Tsinghua University (<http://www.meicmodel>).

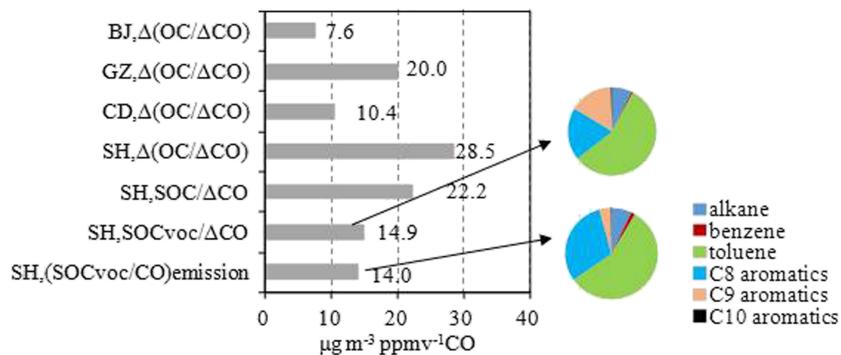


Figure 9. Comparisons of SOC formation in eastern China obtained by different methods. SOC formation in Shanghai (SH, the bottom four columns) based on the VOC emissions inventory (SH, $(\text{SOCvoc/CO})_{\text{emission}}$) and VOC measurements in the ambient environment (SH, SOC_{voc}/CO), total SOC (SH, SOC/ΔCO) estimated by the parameterization method and the increase of the ratio of OC/ΔCO (SH, Δ (OC/ΔCO)) in this study were presented. The contributions of VOC species to SOC formation in Shanghai were shown in color in pie charts. SOC formation in Changdao was indicated by the increase in the ratio of OC/ΔCO (CD, Δ (OC/ΔCO)) (Yuan et al., 2013). SOC formation in Beijing and Guangzhou were indicated by the increase in the ratio of OC/ΔCO (GZ, Δ (OC/ΔCO)) (Hu et al., 2012; Hu et al., 2017).

org/). The speciated VOC emissions inventory in 2013 was obtained from Wu and Xie (2017), and for comparison, only emissions of VOC species measured in the present study were considered. Thus, the ratio of $(SOC_{voc}/CO)_{emission}$ could be calculated from the product of the emission ratios of VOCs to CO, the SOA yields of VOCs under high NO_x levels, and the proportions of VOC photochemical consumption. The proportions of VOC photochemical consumption to total VOC emissions were calculated with the assumption that VOC emissions had the same OH exposure ($\Delta t[OH]$, as shown in equation 2) as the measured VOCs during the pollution episode. Accordingly, the ratio of $(SOC_{voc}/CO)_{emission}$ was estimated to be $14.0 \mu\text{g m}^{-3} \text{ ppmv}^{-1}$, being comparable with the observation-based ratio of SOC_{voc}/CO . However, there were some differences in VOC species contributing to SOC formation between the ambient observation-based result and the emission-based result, as shown in Figure 9. Compared with the observation-based result, the contribution of C9 aromatics to SOC formation was severely underestimated in the emission-based result, and correspondingly the contribution of other VOC species, mainly C8 aromatics, was overestimated. These results suggest that bias might exist in the VOC source profiles in the previous VOC emissions inventory.

4. Conclusions

An episode of heavy photochemical smog was observed in the summer of 2013 in Shanghai, China. During this episode, O_3 and OC in $PM_{2.5}$ increased simultaneously and the VOCs diurnal pattern was completely opposite to those displayed by O_3 and OC. The photochemical processing of VOCs and OC was studied by a parameterization method. VOC photochemical consumption was estimated and found to differ substantially from the ambient measured VOCs both in terms of the diurnal pattern and the species compositions. More than half of the VOCs consumed were aromatics which also dominated (93%) the SOC formation from VOCs. SOC formation from VOCs was estimated to be $3.9 \pm 1.0 \mu\text{g/m}^3$, explaining about 67% of the total SOC ($5.8 \pm 3.6 \mu\text{g/m}^3$) as derived by the parameterization method and further confirmed by the EC tracer method.

According to the OC evolution, the emission ratio of OC to CO was estimated to be $5.8 \pm 0.5 \mu\text{g m}^{-3} \text{ ppmv}^{-1}$, lower than the emissions inventory value of $\sim 8 \mu\text{g m}^{-3} \text{ ppmv}^{-1}$ in Shanghai. Moreover, the ratio of SOC to ΔCO ($SOC/\Delta CO$) was found to be $22.2 \pm 5.7 \mu\text{g m}^{-3} \text{ ppmv}^{-1}$, among which 66.9% came from SOC formation from VOC photochemical processing ($SOC_{voc}/\Delta CO$). For comparison, the ratio of SOC to CO was also calculated based on the emissions inventory of speciated VOCs and CO ($(SOC_{voc}/CO)_{emission}$), assuming that VOC emissions had the same OH exposure as the VOCs measured during the present episode. It turned out that $(SOC_{voc}/CO)_{emission}$ was $14.0 \mu\text{g m}^{-3} \text{ ppmv}^{-1}$ and comparable with the ambient $SOC_{voc}/\Delta CO$, while the precursor compositions were very different. This suggests that there might be a large bias in the VOC source profiles of the emissions inventory.

The present study highlights the importance of the oxidation of VOCs, especially aromatics, for the formation of SOC during photochemical pollution episode in Shanghai. The study also reveals that a large bias likely exists in the species compositions in the emissions inventory. It is essential to improve the VOC emissions inventory, which is critical for making policies and formulating strategies to control air pollution.

Acknowledgments

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