Spatiotemporal variation, sources, and secondary transformation potential of volatile organic compounds in Xi’an, China

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Abstract. As critical precursors of ozone (O₃) and secondary organic aerosols, volatile organic compounds (VOCs) play a vital role in air quality, human health, and climate change. In this study, a campaign of comprehensive field observations and VOC grid sampling was conducted in Xi’an, China, from 20 June to 20 July 2019 to identify the spatiotemporal concentration levels, sources, and secondary transformation potential of VOCs. During the observation period, the average VOC concentrations at the Chanba (CB), Di Huan Suo (DHS), Qinling (QL), and gridded sampling sites were 27.8 ± 8.9, 33.8 ± 10.5, 15.5 ± 5.8, and 29.1 ± 8.4 ppb, respectively. Vehicle exhaust was the primary source of VOC emissions in Xi’an, and the contributions of vehicle exhaust to VOCs at the CB, DHS, and QL sites were 41.3 %, 30.6 %, and 23.6 %–41.4 %, respectively. While industrial emissions were the second-largest source of VOCs in urban areas, contributions from aging sources were high in rural areas. High potential source contribution function values primarily appeared in eastern and southern Xi’an near the sampling site, which indicates that Xi’an exhibits a strong local VOC source. Moreover, alkenes, aromatics, and oxygenated VOCs played a dominant role in secondary transformation, which is a major concern in reducing O₃ pollution in Xi’an.

1 Introduction

Atmospheric pollution in China is characterized by frequent secondary pollution, which is primarily reflected by the yearly increasing ozone (O₃) concentrations and proportion of secondary organic aerosol (SOA) components in PM₂.₅ (Lu et al., 2018; Huang et al., 2014). The increasingly severe secondary pollution has restricted improvements in Chinese ambient air quality. Volatile organic compounds (VOCs) are vital precursors of O₃ and SOA, and reducing VOC emissions is crucial for controlling O₃ and PM₂.₅ pollution (Jin and Holloway, 2015; Yuan et al., 2013).

VOC sources can be divided into biogenic and anthropogenic sources. Although the VOC emissions from biological sources are more than 10 times that of anthropogenic sources globally (Roger and Janet, 2003), anthropogenic sources often dominate the urban atmospheric environment and exert a substantial influence on the formation of O₃ and SOA in the atmosphere (Ahmad et al., 2017; Cuisset et al., 2016; Schwantes et al., 2017). Anthropogenic VOC sources primarily include vehicle exhaust, industrial, paint solvent usage, combustion, and fuel evaporation sources. In recent years, the source apportionment of observation results found that the contribution of vehicle exhaust sources to VOCs plays a leading role in VOC emissions in China, reaching 22 %–58 % (Jia et al., 2016; Liu et al., 2016a, 2020; Zhang et al., 2019). However, the VOC sources exhibit clear differences in various regions. For example, paint solvent usage is the main source of VOCs in Langfang and Xiamen city...
(Zhang et al., 2019; Zhuang et al., 2019), while the primary sources of VOCs in Wuhan city are combustion and vehicle exhaust sources (Shen et al., 2020). The results of VOC emission inventory research indicate that the contribution of vehicle exhaust sources to VOC emissions is gradually decreasing, and industrial sources and paint solvent usage have become the primary sources of VOCs in China (Li et al., 2019; Wu et al., 2016b).

O₃ has become one of the most important pollutants affecting the air quality of the Fenwei Plain. According to the 2018 and 2019 Report on the State of the Ecology and Environment in China released by the Ministry of Ecology and Environment of the People’s Republic of China (http://www.mee.gov.cn/, last access: 24 February 2021), the annual average evaluation value of O₃ concentration in the Fenwei Plain exceeded the Chinese annual evaluation standard (160 µg/m³) for at least 2 consecutive years. Moreover, the average 90th percentile O₃ daily maximum 8 h concentration in the Fenwei Plain was 171 µg/m³ (298.15 K, 1013.25 hPa), which was 4.2 % higher than that in 2018. Xi’an, one of the most famous historical and megacities in Fenwei Plain with a quickly growing economy, has also experienced severe O₃ pollution in recent years. Based on the Report on the State of the Environment in Xi’an (http://xaepb.xa.gov.cn/, last access: 24 February 2021), the average 90th percentile O₃ daily maximum 8 h concentration in Xi’an exceeded the national standards, reaching 169.5, 164.9, and 166 µg/m³ (298.15 K, 1013.25 hPa) in 2017, 2018, and 2019, respectively. In addition, according to statistics released by the Ministry of Public Security, as of the end of 2018, the number of motor vehicles in Xi’an reached 3 million, making Xi’an one of the top eight cities in China in terms of number of motor vehicles. Considering the increasing O₃ concentration and high amount of pollution in Xi’an, studies are imperative in exploring the characteristics, sources, and precursors (VOCs and NOₓ) of O₃ to support the design and implementation of pollution control strategies in Xi’an.

In recent years, numerous VOC observations have been made in many regions of China, particularly in the Beijing–Tianjin–Hebei, Yangtze River Delta, Pearl River Delta, and Chengdu–Chongqing regions. However, there are few studies on Xi’an. Therefore, our understanding of the spatiotemporal characteristics, source contributions, and secondary transformation potential of VOCs in Xi’an remains limited. To elucidate the concentration levels, source characteristics, and secondary conversion ability of O₃ precursors, we conducted comprehensive field observations and VOC grid sampling in Xi’an from 20 June to 20 July 2019. The primary objectives of this study are to (1) determine the temporal and spatial characteristics of atmospheric VOCs in Xi’an, (2) illuminate the characteristics of VOCs during O₃ pollution events, (3) identify critical precursors in Xi’an, and (4) explore VOC sources in Xi’an.

2 Observation and methods

2.1 Observation campaign description

To obtain the temporal and spatial distributions of summertime VOCs in Xi’an, a field observation and VOC grid sampling campaign was applied in this study. The field observation campaign occurred from 20 June to 20 July 2019. Based on an in-person investigation, the prevailing winds, and the electricity supply, three sites were chosen for the field observation campaign (Fig. 1): the Chanba (CB; 34°20′12″ N, 109°01′35″ E), Di Huan Suo (DHS; 34°13′49″ N, 108°52′58″ E), and Qinling (QL; 34°04′11″ N, 108°20′31″ E) sites, which were located on upwind, downtown, and downwind areas of Xi’an city, respectively. The CB site was at the national environmental air quality monitoring urban station (approximately 10 m above ground level) in the Baqiao District, which was close to the third ring road, one of the main traffic thoroughfares in Xi’an. The DHS site was located on the roof of a four-story building (approximately 15 m above ground level) at the Institute of Earth Environment, Chinese Academy of Science in the Yanta District, which was close to the second ring road. The QL site was located in Zhouzhi County, which was adjacent to the Qinling National Botanical Garden. The VOC measurement instrument used in the field observation campaign was an online gas chromatography system that was equipped with a mass spectrometer and flame ionization detector (GC–MS/FID). The GC–MS/FID instrument utilized a dual gas path separation method. The sample gas, after water and CO₂ were removed, captured VOC components through an ultra-low temperature system (−160 °C), and the gas chromatography analysis system was utilized after thermal desorption. The oven temperature was programmed at 37 °C maintained for 5 min initially, then raised to 120 °C at 5 °C min⁻¹ holding for 5 min, and later raised to 180 °C at 6 °C min⁻¹ holding for 5 min. The low carbon number (C₂–C₅) compounds were separated on an Al₂O₃ / KCl PLOT column and quantified using the FID (200 °C). The high carbon number (C₃–C₁₀) compounds were separated on a DB-624 column and quantified using MS (230 °C). External and internal standard gases produced by The Linde Group in the United States were used to calibrate the GC–MS/FID. External standard gas was used to calibrate the GC–MS/FID weekly during the campaign to ensure quantitative accuracy. In addition, the instruments were also daily calibrated by internal standard gases (bromochloromethane, 1,4-dichlorobenzene, chlorobenzene, and fluorobromobenzene) to ensure the stability of the instrument. During the observation period, the GC–MS/FID was in sufficient working condition, and the square correlation coefficients of the VOCs work curves were greater than 0.99. The method detection limits (MDLs) for each VOC compound were calculated according to the TO-15 standard of the United States Environmental Protection Agency (EPA) (Liu et al., 2008b), and the MDL for all measured VOC com-
pounds in this study ranged from 0.002 to 0.121 ppb. O₃, NO, NO₂, and NOₓ concentrations were continuously monitored using i-series chemiluminescence instruments.

VOC gridded samples were collected at each site for two days (1 and 14 July 2019) and twice a day at 07:00 China standard time (CST) and 15:00 (CST). The gridded sampling site was chosen based on the technical regulations for selecting ambient air quality monitoring stations (HJ 664–2013) and method for selection of Photochemical Assessment Monitoring Stations (EPA). According to the prevailing wind direction and to ensure coverage of all urban areas in Xi’an, 20 gridded sampling sites were selected for this study (Fig. 1). Detailed sampling site information is shown in Table 1. A total of 80 ambient air samples with a frequency of 4 samples per site were collected, and each sample was stored in a 3.2 L Silonite™ canister (Entech Instrument, United States). Before VOC gridded sampling, the Silonite™ canisters were cleaned with high-purity nitrogen using the Entech 3100 canister cleaning system, and then they were evacuated to a vacuum. The instantaneous sampling method was adopted for ambient air sample collection with a sampling duration of approximately 2 min. VOCs in the sampled air were analyzed using a GC–MS/FID system, which was the same as that used for online measurements, but it was running in offline mode. In this study, blank sample tests were performed on the VOC gridded sampling in each sampling period (1 July 07:00, 1 July 15:00, 14 July 07:00, and 14 July 15:00). The concentration of all VOC species in the blank sample is below the MDLs, indicating that the canisters were not contaminated during transportation.

2.2 Photochemical reactive activity parameterization

The loss rates of VOCs that react with OH radicals (L_OH) and the O₂ formation potential of OFP) can be used to characterize VOC photochemical activity (Carter, 2010; Niu et al., 2016; Wu et al., 2016a). L_OH and OFP can be calculated using Eqs. (1) and (2), respectively.

\[
L_{\text{OH}} = \sum_{i} n_{\text{O}_{\text{CH}}} \times [\text{VOC} \text{ (molec. cm}^{-3})], \tag{1}
\]

\[
\text{OFP} = \sum_{i} n_{\text{MIR}_{i}} \times [\text{VOC (ppb)}], \tag{2}
\]

where \(n\) represents the number of VOCs, \([\text{VOC}]\) represents the \(i\)th VOC species concentration, \(k_{\text{O}_{\text{CH}}}\) represents the rate coefficient for the reaction of the \(i\)th VOC species with OH radical (molec. \text{cm}^{-3} \text{s}^{-1}), \text{and MIR}_{i}\) is the maximum incremental reactivity for the \(i\)th VOC species. The \(k_{\text{O}_{\text{CH}}}\) and MIR for each VOC species were taken from the updated Carter research results (http://www.engr.ucr.edu/~carter/reactdat.htm, last access: 24 February 2021).

Specific VOC ratios are often used to calculate the air mass photochemical age or OH exposure (Jimenez et al., 2009; Roberts et al., 1984). OH exposure can be calculated using Eq. (3).

\[
\text{OH exposure} = \frac{1}{(k_{\text{voc1}} - k_{\text{voc2}})} \times \left[ \ln \frac{[\text{voc1}]}{[\text{voc2}]} \right]_{t=0} - \ln \frac{[\text{voc1}]}{[\text{voc2}]}, \tag{3}
\]

where OH exposure represents OH radical concentration multiplied by the reaction time (\(\Delta t\)); \(k_{\text{voc1}}\) and \(k_{\text{voc2}}\) represent rate coefficients for the reaction of the VOC species with OH radical (molec. \text{cm}^{-3} \text{s}^{-1}); \([\text{voc1}]\) represents the initial emission ratio of specific VOCs, which can be replaced by the highest concentration ratio in periods where the photochemical reaction is weak; and \([\text{voc2}]\) represents the concentration ratio of the specific VOCs in the atmosphere.

2.3 Positive matrix factorization (PMF)

The PMF analysis model was first proposed by Paatero and Tapper (1994). For more than two decades, PMF has been widely used to identify and quantify major sources of VOCs (He et al., 2019; Li et al., 2015; Miller et al., 2002; Pallavi et al., 2019; Song et al., 2007; Yuan et al., 2010). The definition and usage of the PMF model is described elsewhere in detail (Liu et al., 2020; Song et al., 2018; Su et al., 2019), and only a brief description is provided here.

In this study, the PMF 5.0 model (EPA) was used to analyze the VOC sources at the CB, DHS, and QL sites. VOCs tracers were selected according to the reported typical emission source profiles in China (He et al., 2015; Liu et al., 2008a; Song et al., 2018). Only those tracers which have data coverage greater than 75% during the campaign and have 65% measured concentrations above the MDL were included in the PMF analysis (cf. Liu et al., 2020). In this study, the number of VOC tracers (input species) at the CB, DHS, and QL sites were 32, 34, and 32, respectively. A concentration (Conc.) file and uncertainty (Unc) file are required by PMF. For the concentration file, data below the detection limit were assigned with MDL/2, and the missing data were substituted with median concentration. The uncertainty is calculated using Eq. (4) as follows (Brown et al., 2015; Liu et al., 2016a):

\[
\text{Unc} = \begin{cases} 
\frac{\sqrt{\text{Error Fraction} \times \text{Conc.}^2 + (0.5 \times \text{MDL})^2}}{\text{Conc.} > \text{MDL}}, \\
\frac{\text{Conc.} \leq \text{MDL}}{4 \times \text{median conc.}}, \\
\text{Missing data}
\end{cases}
\]  

In this study, the PMF factor numbers were explored from 4 to 8 for the optimal solution in the three sites. \(Q_{\text{true}} / Q_{\text{robust}}\) and \(Q_{\text{true}} / Q_{\text{expected}}\) are two important parameters for characterizing the rationality of the PMF results (Brown et al., 2015). After comparing the PMF results, \(Q_{\text{true}} / Q_{\text{robust}}\) ratio, and \(Q_{\text{true}} / Q_{\text{expected}}\) ratio, a seven-factor PMF solution was selected for VOC source apportionments in the three field observation sites. The \(Q_{\text{true}} / Q_{\text{robust}}\) values at the CB, DHS, and QL sites were all 1.0. The \(Q_{\text{true}} / Q_{\text{expected}}\) values at the CB, DHS, and QL sites were 1.3, 1.2, and 1.0, respectively. In addition, an \(F_{\text{peak}}\) parameter, from −1.0 to 1.0 (step of 0.1), was used to rotate the PMF factors for a superior solution (Sun et al., 2012). In this study, the factor rotating results

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Figure 1. VOC field observation and grid sampling sites in Xi’an (© Google Earth). The topographic image was obtained from Google Earth.

Table 1. Detailed sampling site information.

<table>
<thead>
<tr>
<th>Site name</th>
<th>Abbreviation</th>
<th>Latitude</th>
<th>Longitude</th>
<th>Parameter</th>
</tr>
</thead>
<tbody>
<tr>
<td>Chanba</td>
<td>CB</td>
<td>34°20'12'' N</td>
<td>109°01'35'' E</td>
<td>O3/NO/NO2/NOx/CO/photolysis rate/VOCs (99)/T/P/RH</td>
</tr>
<tr>
<td>Di Huan Suo</td>
<td>DHS</td>
<td>34°13'49'' N</td>
<td>108°52'58'' E</td>
<td></td>
</tr>
<tr>
<td>Qinling</td>
<td>QL</td>
<td>34°04'11'' N</td>
<td>108°20'31'' E</td>
<td></td>
</tr>
<tr>
<td>Caotan</td>
<td>CT</td>
<td>34°23'4'' N</td>
<td>108°52'13'' E</td>
<td>VOCs (106)</td>
</tr>
<tr>
<td>College of Automotive Technology</td>
<td>CAT</td>
<td>34°12'48'' N</td>
<td>109°5'54'' E</td>
<td></td>
</tr>
<tr>
<td>Gengzheng Subdistrict Office</td>
<td>GZ</td>
<td>34°26'52'' N</td>
<td>109°7'31'' E</td>
<td></td>
</tr>
<tr>
<td>Guang Yun Lake</td>
<td>GYL</td>
<td>34°19'23'' N</td>
<td>109°3'1'' E</td>
<td></td>
</tr>
<tr>
<td>Hu County</td>
<td>HC</td>
<td>34°6'43'' N</td>
<td>108°30'36'' E</td>
<td></td>
</tr>
<tr>
<td>Jiaodai Town</td>
<td>JDT</td>
<td>34°3'33'' N</td>
<td>109°14'39'' E</td>
<td></td>
</tr>
<tr>
<td>Juien Town</td>
<td>JFT</td>
<td>34°3'35'' N</td>
<td>108°25'50'' E</td>
<td></td>
</tr>
<tr>
<td>Lantian County</td>
<td>LTC</td>
<td>34°9'9'' N</td>
<td>109°19'5'' E</td>
<td></td>
</tr>
<tr>
<td>Luan Town</td>
<td>LT</td>
<td>33°59'14'' N</td>
<td>108°50'4'' E</td>
<td></td>
</tr>
<tr>
<td>Region Site</td>
<td>RS</td>
<td>34°10'45'' N</td>
<td>108°44'18'' E</td>
<td></td>
</tr>
<tr>
<td>Wangqu Subdistrict Office</td>
<td>WQ</td>
<td>34°4'36'' N</td>
<td>108°59'54'' E</td>
<td></td>
</tr>
<tr>
<td>Xiaoyan</td>
<td>XY</td>
<td>34°13'4'' N</td>
<td>108°56'15'' E</td>
<td></td>
</tr>
<tr>
<td>Xiehu Town</td>
<td>XHT</td>
<td>34°13'47'' N</td>
<td>109°13'41'' E</td>
<td></td>
</tr>
<tr>
<td>Xinfeng Subdistrict Office</td>
<td>XF</td>
<td>34°25'23'' N</td>
<td>109°16'12'' E</td>
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<tr>
<td>Xinshi Subdistrict Office</td>
<td>XS</td>
<td>34°30'38'' N</td>
<td>109°11'7'' E</td>
<td></td>
</tr>
<tr>
<td>Xinxing Subdistrict Office</td>
<td>XX</td>
<td>34°39'43'' N</td>
<td>109°16'29'' E</td>
<td></td>
</tr>
<tr>
<td>Yin Town</td>
<td>YT</td>
<td>34°0'17'' N</td>
<td>109°6'44'' E</td>
<td></td>
</tr>
<tr>
<td>Yushan Town</td>
<td>YST</td>
<td>34°12'53'' N</td>
<td>109°31'33'' E</td>
<td></td>
</tr>
<tr>
<td>Zhouzhi County</td>
<td>ZZC</td>
<td>34°9'35'' N</td>
<td>108°12'33'' E</td>
<td></td>
</tr>
<tr>
<td>Zhuyu Town</td>
<td>ZYT</td>
<td>34°5'50'' N</td>
<td>108°7'41'' E</td>
<td></td>
</tr>
</tbody>
</table>

T: temperature; P: pressure; RH: humidity.
were not significantly different than the non-rotation results. Thus, the results used in this study were from the runs with zero Fpeak.

2.4 Conditional probability function (CPF) analyses

A CPF analysis can be used to explore the PMF-identified VOC source impacts at varying wind directions and speeds (Huang and Hsieh, 2019; Liu et al., 2016a). The CPF can be calculated using Eq. (5) as follows (Ashbaugh et al., 1985; Uria-Tellaetxe and Carslaw, 2014):

\[
CPF = \frac{m_{\theta, \mu, 75\text{per}}}{n_{\theta, \mu}},
\]

where \(m_{\theta, \mu, 75\text{per}}\) is the number of samples in the wind direction \(\theta\) and wind speed interval \(\mu\) with a VOC concentration greater than the 75th percentile concentration, and \(n_{\theta, \mu}\) is the total number of samples in the same wind direction and speed interval.

2.5 Cluster and potential source contribution function (PSCF) analysis

Cluster analyses of backward trajectories are widely used to determine a dominant air mass direction and potential origin direction of pollutants at the study sites (Hong et al., 2019; Liu et al., 2016a, 2019). In this study, the 24 h backward trajectories (1 h intervals) of air masses arriving 100 m above ground level were calculated using the MeteoInfoMap software. Air mass reanalysis data were obtained from the National Weather Service’s National Centers for Environmental Prediction at a gridded resolution of 0.25° × 0.25° (ftp://arlftp.arlhq.noaa.gov/pub/archives/gfs0p25, last access: 24 February 2021). The Euclidean distance clustering algorithm was chosen to cluster the air mass trajectories in the CB, DHS, and QL sites from 20 June to 20 July 2019. To obtain the optimal clustering solution, the clustering result with the smallest percent change was selected (Wang et al., 2010).

PSCF is a gridded statistical analysis method that is used to combine the backward trajectory and the corresponding VOC concentration to determine the potential VOC source area (Liu et al., 2016b, 2020; Zheng et al., 2018). In this study, Xi’an city and the adjacent region covered by the back trajectories were divided into an array of 0.25° × 0.25° grid cells. In this study, the pollution trajectory was defined as the trajectories corresponding to the total VOC (TVOC) concentration that exceeded the 75th percentile concentration of TVOCs. When the number of endpoints of the pollution trajectory passing through the grid \((i, j)\) is \(M_{ij}\), and the number of endpoints of all the trajectories falling within the grid \((i, j)\) is \(N_{ij}\), then PSCF\(_{ij}\) can be defined as the ratio of \(M_{ij}\) to \(N_{ij}\) (Polissar et al., 1999). The weight function \(W_{ij}\) was used to increase the accuracy of the model, and PSCF\(_{ij}\) can be calculated using Eq. (6) as follows:

\[
PSCF_{ij} = M_{ij} \times \frac{W_{ij}}{N_{ij}} = \frac{M_{ij}}{N_{ij}} \times \begin{cases} 1.00 & N_{ij} > 80 \\ 0.70 & 40 < N_{ij} \leq 80 \\ 0.42 & 5 < N_{ij} \leq 40 \\ 0.05 & N_{ij} \leq 5 \end{cases}
\]

2.6 Empirical kinetic modeling approach

The traditional empirical kinetic modeling approach (EKMA) is a model sensitivity test of the observation-based box model and often used to evaluate the photochemical nonlinear relationship between ozone and precursors NO\(_x\) and VOCs. The box model is based on the Regional Atmospheric Chemical Mechanisms version 2 (Goliff et al., 2013) updated with the Leuven Isoprene Mechanism (Peeters et al., 2009). The definition and mechanism of the observation-based box model is described elsewhere in detail (Tan et al., 2017, 2018).

The EKMA curve can be used as a theoretical basis for designing emission reduction strategies to obtain the best ozone pollution reduction method (Jiang et al. 2018; Tan et al., 2018). The model input parameters include temperature, pressure, humidity, photolysis rate constant, NO\(_x\), and VOCs. In this model, the ozone production rate \(P\) (O\(_3\)) can be calculated using Eq. (7) as follows:

\[
P_{O_3} = F_{O_3} - D_{O_3},
\]

\[
F_{O_3} = k_{RO_2+NO}[HO_2][NO] + k_{(RO_2+NO)_{\text{eff}}}[RO_2][NO],
\]

\[
D_{O_3} = \left[O^{1D}\right][H_2O] + k_{O_3+OH}[O_3][OH] + k_{O_3+HO_2}[O_3][HO_2] + k_{O_3+alkenes}[O_3][alkenes] + k_{OH+NO_2}[OH][NO_2],
\]

where \(k_{RO_2+NO}\) represents rate coefficients for the reaction of the NO with HO\(_2\) radical (8.5 \(\times\) 10\(^{-12}\) molec.\(^{-1}\) cm\(^3\) s\(^{-1}\), 298 K), \(k_{(RO_2+NO)_{\text{eff}}}\) represents effective rate coefficients for the reaction of the NO with RO\(_2\) radical (8.5 \(\times\) 10\(^{-12}\) molec.\(^{-1}\) cm\(^3\) s\(^{-1}\), 298 K), \(k_{OH+NO_2}\) represents rate coefficients for the reaction of the O\(_3\) with OH radical (7.3 \(\times\) 10\(^{-14}\) molec.\(^{-1}\) cm\(^3\) s\(^{-1}\), 298 K), \(k_{O_3+alkenes}\) represents rate coefficients for the reaction of the O\(_3\) with alkenes (1.9 \(\times\) 10\(^{-15}\) molec.\(^{-1}\) cm\(^3\) s\(^{-1}\), 298 K), \(k_{O_3+alkenes}\) represents rate coefficients for the reaction of the O\(_3\) with alkenes (2.0 \(\times\) 10\(^{-17}\) molec.\(^{-1}\) cm\(^3\) s\(^{-1}\), 298 K), and \(k_{OH+NO_2}\) represents rate coefficients for the reaction of the NO\(_2\) with OH radical (1.1 \(\times\) 10\(^{-11}\) molec.\(^{-1}\) cm\(^3\) s\(^{-1}\), 298 K).

In this study, the average parameters of the entire observation period were used as the input parameters of the model to calculate the ozone concentration in the baseline scenario. Afterwards, the activity change array of VOCs and NO\(_x\) were generated by changing each parameter in equal distance.
steps. The $P$ ($O_3$) contours under these different VOCs and NO$_x$ reactivity conditions are called EKMA curves.

3 Results and discussion

3.1 Temporal and spatial characteristics of VOCs

3.1.1 Temporal variations

Temporal variations in wind speed, wind direction, temperature, relative humidity, $O_3$, and VOCs at the CB, DHS, and QL sites are shown in Fig. 2. During the field observation campaign, 99 VOCs were measured, including 29 alkanes, 11 alkenes, 1 alkyn, 16 aromatics, 28 halohydrocarbons, 13 oxygenated VOCs (OVOCs), and 1 acetonitrile (Supplement Table S1). The CB and DHS sites were located in an urban area of Xi’an, while the QL site was located in a rural area. During the observation period, the average VOC concentrations at the CB, DHS, and QL sites were 27.8 ± 8.9, 33.8 ± 10.5, and 15.5 ± 5.8 ppb, respectively. Due to the existence of more emission sources and lower wind speeds (0–2 m/s) in urban areas, the VOC concentrations were higher than those in rural areas. Overall, the VOC concentrations in Xi’an urban sites were approximately twice that of the rural sites. The observation period occurred during summer in Xi’an, and the temperature was high, with average temperatures reaching 26.2 ± 4.3 and 25.6 ± 3.9°C in urban and rural areas, respectively. Higher temperatures may accelerate a secondary transformation of VOCs into $O_3$, resulting in more frequent $O_3$ pollution incidents in Xi’an; the mean $O_3$ concentrations at the CB, DHS, and QL sites reached 50.2 ± 29.9, 47.6 ± 29.4, and 19.7 ± 8.0 ppb, respectively.

During the observation period, there were two $O_3$ pollution events in the Xi’an urban area (CB and DHS sites), from 23 to 26 June 2019 and from 30 June to 15 July 2019. The wind speed diurnal pattern in the urban area displayed a single wave profile on both clean and polluted days with low wind speeds (0.4 ± 0.2 and 0.3 ± 0.1 m/s, respectively) during the night and peaks (0.7 ± 0.2 and 0.9 ± 0.3 m/s, respectively) at 14:00 CST. The overall wind speed in the Xi’an urban area was in a static state, indicating that the ability of wind to diffuse and dilute pollutants is limited (Fig. 3).

The variation trends of $O_3$ and temperature display a positive correlation, and the linear correlations between $O_3$ and temperature on polluted days ($R_{\text{Pearson}} = 0.7$) is stronger than that on clean days ($R_{\text{Pearson}} = 0.5$). The value of temperature, $O_3$, and TVOCs all increased significantly on polluted days, indicating that the secondary transformation of VOCs to $O_3$ is more conducive at high temperatures. The difference between the VOC species concentration on clean and polluted days at the CB and DHS sites is shown in Fig. 3c and e. As shown in Fig. 3, isoprene concentrations at urban sites increased significantly during the $O_3$ pollution day, which could be due to the stronger plant emission at elevated temperature (Guenther et al., 1993, 2012; Stavrakou et al., 2014). Concentrations of isoprene oxidation products (i.e., MVK and MACR) as well as most OVOCs also increased in the same period. However, similar concentrations of anthropogenic VOCs are found in clean and polluted days. This indicates a stronger photochemical conversion of VOCs existed in $O_3$ pollution days, which could due to the more favorable meteorological conditions (i.e., higher temperature and solar radiation). The specialty of OVOCs is that in addition to the primary emissions, OVOCs can also be formed through photochemical oxidation with alkenes and aromatics (Birdsall and Elrod, 2011). The sources of OVOCs can be divided into anthropogenic primary sources, anthropogenic secondary sources, biogenic sources, and background sources (Li et al., 2014; Wang et al., 2015). Based on the multi-linear regression model results (Figs. S1 and S2) we found that the contribution of anthropogenic primary sources to OVOCs on $O_3$ pollution days is more significant.

3.1.2 Spatial variations

In this study, the VOC grid sampling was used to investigate the spatial variations in VOCs in Xi’an. A total of 20 sites were selected for grid monitoring that covered the entire city of Xi’an. Therefore, the results of the VOC grid sampling were used to represent the levels of the entire city. In the VOC grid sampling, 106 VOCs were measured, including 29 alkanes, 11 alkenes, 1 alkyn, 17 aromatics, 35 halohydrocarbons, 12 OVOCs, and carbon disulfide (Table S2-S3). The weather parameter on 1 and 14 July in Xi’an is shown in Table S4. The average VOC concentration at the 20 sites on 1 July and 14 July 2019 was 29.1 ± 8.4 ppb, and high VOC concentrations were clustered at the XF, CT, HC, ZYT, and YT sites (Fig. 4a and b). Of the sites, the XF site exhibited the highest VOC concentration of 54.0 ppb, followed by CT and HC, with concentrations of 41.4 and 38.2 ppb, respectively. The XHT site exhibited the lowest concentration of VOCs at 18.0 ppb, followed by WQ and YST sites with concentrations of 19.6 and 19.9 ppb, respectively. Compared with the results of the field observation campaign, the VOC concentration at the CB site was closer to the average concentration in Xi’an, and the VOC concentration at the DHS site was significantly higher than the average concentration. In terms of the VOC composition at each site, alkanes and OVOCs were dominant, accounting for 25.7%–39.7% and 22.8%–47.4% of the VOC concentration, respectively. In addition, the contribution of OVOCs at the YT site was significantly higher than that of the other sites, indicating that the YT site may be significantly affected by aging sources (Fig. 4a).

The top 10 VOC species at the CB, DHS, QL, and grid sampling sites accounted for 66.1%, 63.4%, 71.1%, and 67.1% of the TVOC concentration, respectively (Table 2). Of these species, ethane, acetone, and propane were the top three contributors in Xi’an, accounting for 34.0%–41.3% of the TVOC.
3.2 Sources

3.2.1 Specific VOC ratios

Different VOC species may have different sources; hence, the ratio of different species can be used to preliminarily analyze the difference in VOC sources at each site. The ratios of specific species that are often used are the toluene/benzene (T/B), m/p-xylene/ethylbenzene (X/E), iso-pentane/n-pentane, benzene/ethyne, toluene/ethyne, and m/p-xylene/ethyne ratios.

The T/B ratio is clearly different for various source profiles (Table S5). In industrial source emission studies, the ratio of T/B ranged from 1.4 ± 0.8 to 5.8 ± 3.4 by different industry type and process unit (Mo et al., 2015; Shi et al., 2015). In traffic source emission studies, the ratio of T/B ranged from 0.9 ± 0.6 to 2.2 ± 0.5 by different vehicle type and fuel composition (Qiao et al., 2012; Dai et al., 2013; Wang et al., 2013; Yao et al., 2013; Zhang et al., 2013; Yao
et al., 2015b; Mo et al., 2016; Yao et al., 2015a; Deng et al., 2018). However, vehicle emissions include both diesel vehicles emissions and gasoline vehicles emissions in the atmospheric environment. Thus, the ratio of T/B in the traffic source should be closer to the results of the tunnel experiments which are approximately 1.5 ± 0.1 (Liu et al., 2008a; Deng et al., 2018). In paint solvent usage source emission studies, the ratio of T/B was greater than 8.8 ± 6.5 by different solvent use process (Yuan et al., 2010; Wang et al., 2014; Zheng et al., 2013). In burning source emission studies, the T/B ratio was approximately 0.3 ± 0.1 in different combustion process and raw materials (Liu et al., 2008a; Li et al., 2011; Mo et al., 2016). In order to reduce the influence of photochemical reaction on the ratio of benzene to toluene, this study selected the weaker photochemical reaction period (03:00–07:00) for the analysis of toluene and benzene (Fig. 5). Figure 5 shows that the ratios of toluene to benzene at the CB, DHS, QL, and gridded sampling sites were 1.1 (R\_Pearson = 0.5), 3.6 (R\_Pearson = 0.6), 0.5 (R\_Pearson = 0.8), and 1.75 (R\_Pearson = 0.9), respectively. In the urban areas (CB and DHS sites), most of the T/B ratios were distributed within the reference range of vehicle emissions and industrial emissions (Fig. 5a, b), implying that vehicle sources and industrial sources contribute significantly to the VOCs in the Xi’an urban area. In addition, the T/B value of some samples is greater than 5.8 in the urban area, which may be affected by paint solvent usage source (Fig. 5b). However, the detailed source contribution needs to be obtained through PMF source analysis results (Sect. 3.2.2). In the rural area (QL site), most of the T/B ratios were distributed within the reference range of vehicle emissions and burning emissions (Fig. 5c), implying that vehicle sources and burning sources contribute significantly to the VOCs in the Xi’an rural area. In the gridded sampling sites, the T/B ratio was predominately concentrated around 1.5, indicating that vehicle exhaust sources may greatly contributed to the overall VOCs in Xi’an (Fig. 5d).

The reactivity of m/p-xylene with OH radicals was 2.7 times that of ethylbenzene (Carter, 2010); therefore, a lower X/E ratio represents a higher degree of air mass aging. In areas with high air mass aging, the contribution of external source transport to VOCs increased significantly. The diurnal variation of the X/E ratios at the CB, DHS, and QL sites demonstrates that the X/E ratio at the three sites significantly decreased from 09:00 to 13:00 (Fig. 6a–c). This indicated that there was a significant photochemical consumption effect on VOCs between 09:00 and 13:00. In addition, the X/E ratios at the CB and DHS sites significantly increased after 13:00, while variations in the X/E ratio were not clear at the QL site, indicating that there were more primary emissions from anthropogenic sources at the urban sites (CB and DHS sites). The diurnal variations in OH exposure exhibit an inverse correlation with the X/E ratio, reaching a maximum daily value between 12:00–15:00 (Fig. 6d–f). The average OH exposures of the CB, DHS, and QL sites were 5.1 \times 10^{10}, 1.7 \times 10^{10}, and 3.1 \times 10^{10} molec. cm^{-3} s, respectively.

The ratio of iso-pentane/n-pentane is clearly different for various sources. Recent studies have found that the iso-pentane/n-pentane ratio was 2.93 for vehicle ex-
Figure 6. Diurnal variations in \( m/p \)-xylene to ethylbenzene and OH exposure at the CB, DHS, QL, and gridded sampling sites. Time is expressed in CST. The green line represents the initial emission ratio of \( m/p \)-xylene and ethylbenzene.

Figure 7. Linear correlations \( R^2 \) between (a) iso-pentane and \( n \)-pentane and (b) propane and ethane at the CB (red), DHS (orange), QL (blue), and gridded sampling sites (light blue).

3.2.2 Sources apportionment

During the observation period (20 June to 20 July 2019), seven-factor PMF solutions were selected for the CB, DHS, and QL sites, and nine possible emission sources were identified: vehicle exhaust, industrial sources, combustion, paint solvent usage, biogenic sources, fuel vaporization, biogenic burning, aging sources, and vehicle exhaust + industrial sources. The source profiles and contributions to VOC concentrations in the CB, DHS, and QL sites during the observation period are displayed in Fig. 8. The relationships between wind direction, wind speed, and the main sources of the three stations are illustrated in Fig. 9.

Vehicle exhaust sources are characterized by high concentrations of alkanes (Cai et al., 2010), a few alkenes (An et al., 2017; Liu et al., 2008a), ethyne (Ling et al., 2011), and \( C_6 - C_7 \) aromatics (Mo et al., 2018). In this study, vehicle exhaust sources in the CB and QL sites exhibited high ethane, propane, butane, and acetone contents. In the DHS site, vehicle exhaust sources exhibited high ethane, propane, butane, pentane, ethylene, ethyne, and acetone contents and a few aromatics. These species are important vehicle exhaust tracers. Thus, these factors are identified as sources of vehicle exhaust. At the end of 2019, the number of motor vehicles in Xi’an city exceeded 3.43 million, with the city ranking eighth in China for motor vehicle number (https://www.mps.gov.cn, last access: 24 February 2021). Moreover, the contribution of
Vehicle exhaust to VOCs at the CB, DHS, and QL sites was 41.3%, 30.6%, and 23.6%–41.4% (Fig. 8), respectively. Vehicle exhaust has become an important source of VOC emissions in Xi’an. The CPF results in Fig. 9 indicated that vehicle exhaust at the CB site had a high potential (CPF > 0.6) of source transport from the southwest when the wind speed exceeded 1 m/s. However, the vehicle exhaust at the DHS and QL sites was not significantly affected by the transmission source (CPF < 0.3).

Industrial sources are characterized by high concentrations of halohydrocarbons (Dumanoglu et al., 2014; Sun et al., 2016) and aromatics (Guo et al., 2011). They also contain alkanes and alkenes, such as ethane, hexane, and ethylene (Liu et al., 2008a), which are important raw materials for industrial production. Factors that meet these characteristics were identified in this study as industrial sources. The industrial sources of the CB site contained a higher proportion of \(n\)-undecane and \(n\)-dodecane, which are important tracers for asphalt industry applications (Liu et al., 2008a). The source contribution results demonstrate that industrial emissions are the second-largest source of VOC emissions in the urban areas of Xi’an, accounting for 29.7% and 30.6% at the CB and DHS sites, respectively (Fig. 8). Because the QL site is located in a rural area with few factories, the contribution of industrial sources is small. The CPF results (Fig. 9) indicated that industrial sources at the CB and DHS sites were not significantly affected by the transmission source (CPF < 0.3).

Combustion sources are characterized by high concentrations of ethane, ethylene, propane, and ethyne (Ling et al., 2011; Liu et al., 2008a; Song et al., 2018). Factors that meet these characteristics were identified as combustion sources in this study. Combustion sources have become the third-largest source of VOCs in Xi’an, accounting for 10.9%, 12.3%, and 12.8% at the CB, DHS, and QL sites, respectively (Fig. 8). The CPF results demonstrated that combustion sources at the QL site exhibited a high potential (CPF > 0.8) of source transport from the north when the wind speed exceeded 3 m/s (Fig. 9).

Paint solvent usage sources are characterized by high concentrations of aromatics, such as toluene, ethylbenzene, \(m/p\)-xylene, and \(o\)-xylene (An et al., 2017; Li et al., 2018; Song et al., 2019). Factors that meet these characteristics were identified as paint solvent usage sources in this study. Paint solvent usage sources contributed significantly to the VOCs in Xi’an, accounting for 8.2%, 12.1%, and 11.2% at
Figure 9. Polar plot of primary VOC sources in (a) CB, (b) DHS, and (c) QL sites based on the CPF function.

The CB, DHS, and QL sites, respectively (Fig. 8), which was comparable to the contribution of combustion sources.

Biogenic sources are characterized by high concentrations of isoprene and the oxidation products of isoprene (methacrolein and methyl vinyl ketone) (Gong et al., 2018; Ling and Guo, 2014; Ling et al., 2019). Factors that meet these characteristics were identified as biogenic sources in this study. Biogenic sources were identified at the DHS and CB sites, accounting for 7.2% in both sites (Fig. 8).

Biogenic burning sources are characterized by high concentrations of ethyne, methyl chloride, benzene, and toluene (Liu et al., 2008a). The fifth factor of the QL site met this characteristic and was identified as a biogenic burning source, accounting for 8.2% (Fig. 8).

Fuel evaporation sources are characterized by high concentrations of iso-butane, n-butane, iso-pentane, n-pentane, 2-methylpentane, and 3-methylpentane (Liu et al., 2017; Zheng et al., 2020). The seventh factor of the CB site met this characteristic and is identified as a fuel evaporation source, accounting for 9.8% (Fig. 8).

Aging sources are characterized by high concentrations of OVOCs (Li et al., 2015; Zhu et al., 2018). As important tracers of aging sources, OVOCs include both primary and secondary sources and have a longer lifetime in the atmosphere (Derstroff et al., 2017). The fourth factor of the QL site met this characteristic and was identified as an aging source, accounting for 19.2% (Fig. 8). The CPF results illustrate that aging sources at the QL site exhibited a high potential (CPF > 0.8) of source transport from the east when the wind speed exceeded 1 m/s (Fig. 9).

3.3 Cluster and PSCF results

The 24 h backward trajectories from Xi’an for the cluster and PSCF analysis are shown in Fig. 10. Based on the figure, air mass back trajectories can be clustered into the eastern trajectories (Cluster 1), southeastern trajectories (Cluster 2 + 4), and northeastern trajectories (Cluster 3 + 5) at the CB site (Fig. 10a). It is evident that the proportion of southeastern trajectories to the total trajectories and that of the southeastern pollution trajectories to the total pollution trajectories were significantly higher than those of the other cluster trajectories, accounting for 58.7% and 60.8%, respectively (Table 3). There were two trajectory clusters from the southeast direction, the southeast short distance trajectories (Cluster 2) and southeast medium-long-distance trajectories (Cluster 4), accounting for 35.2% and 23.5%, respectively. This result indicated that the VOC concentration in the
CB site was significantly affected by the southeast trajectory from the junction of the Shaanxi Province, Hubei Province, and Henan Province in addition to local sources. In addition, although the proportion of pollution trajectories from the northwest (LDT) was small (Cluster 3), the concentrations in these pollution trajectories were the greatest, reaching 41.5 ppb (Table 3). Thus, attention should be paid to the long-distance transmission of highly polluting air masses from Inner Mongolia. Regarding the VOC composition (Fig. 10b), alkanes with lower activities accounted for a larger proportion in the long-distance trajectories (Cluster 3).

The air mass back trajectories can be clustered into the eastern trajectories (Cluster 1), southeastern trajectories (Cluster 2 + 4), and northeastern trajectories (Cluster 3 + 5) at the DHS site (Fig. 10c). It is evident that the proportion of southeast trajectories to the total trajectories and that of the southeast pollution trajectories to the total pollution trajectories were significantly higher than those of the other cluster trajectories, accounting for 68.9 % and 73.3 %, respectively (Table 3). This result indicated that the VOC concentration in the DHS site was mainly affected by pollution transmission at the Shaanxi Province, Hubei Province, and Henan Province junction and local source emissions. In addition, although the proportion of pollution trajectories from the northwest (LDT) was small (Cluster 3), the concentrations in these pollution trajectories were the greatest, reaching 57.9 ppb. Thus, attention should be paid to the long-distance transmission of highly polluting air masses from central Inner Mongolia.

The air mass back trajectories can be clustered into the eastern trajectories (Cluster 1), northeastern trajectories (Cluster 2), southeastern trajectories (Cluster 3 + 4), and northeastern trajectories (Cluster 5) at the QL site (Fig. 10e). It is evident that the proportion of northeastern trajectories to the total trajectories and that of the northeastern pollution trajectories to the total pollution trajectories were significantly higher than those of the other cluster trajectories, accounting for 36.4 % and 34.3 %, respectively (Table 3). However, the VOC concentrations of the eastern trajectories to the total trajectories were significantly higher than those of the other cluster trajectories, accounting for 44.2 %–54.5 %.

In the eastern and southern regions of the site, the highest PSCF values of TVOCs appeared in areas near the CB, DHS, and QL sites, which indicated that Xi’an has a strong local source.

### 3.4 Secondary transformation potential

#### 3.4.1 L\textsubscript{OH} and OFP

The VOC loss rates (L\textsubscript{OH}) that react with OH at the CB, DHS, and QL sites were 4.9, 4.8, and 2.4 s\textsuperscript{-1}, respectively. The average L\textsubscript{OH} of the VOCs at the 20 gridded sites was 5.0 s\textsuperscript{-1}, and high VOC L\textsubscript{OH} was concentrated at the XF, ZYT, CT, LTC, and LT sites (Fig. 11a). Of the sites, XF exhibited the highest L\textsubscript{OH} of 16.2 s\textsuperscript{-1}, followed by ZYT and CT with L\textsubscript{OH} of 9.9 and 6.6 s\textsuperscript{-1}, respectively. The XX site exhibited the lowest L\textsubscript{OH} of 1.6 s\textsuperscript{-1}, followed by the WQ and YST sites with L\textsubscript{OH} of 1.6 and 2.3 s\textsuperscript{-1}, respectively. At the CB, DHS, and QL sites, alkenes and OVOCs played a dominant role in L\textsubscript{OH}, accounting for 31.7 %–50.8 % and 26 %–35.8 % (Fig. 11a). For the gridded sampling sites, the overall L\textsubscript{OH} in Xi’an was significantly affected by alkenes, accounting for 45.4 %–85.9 %. The top 10 VOC species of L\textsubscript{OH} at the CB, DHS, QL, and grid sampling sites accounted for 69.2 %, 60.8 %, 75.9 %, and 80.2 % of the total L\textsubscript{OH}, respectively (Table 2). Of these species, isoprene, acetaldehyde, ethylene, and 1-butene were the greatest contributors to total L\textsubscript{OH} in Xi’an, reaching 36.4 %–63.0 %.

The O\textsubscript{3} formation potentials of VOCs at the CB, DHS, and QL sites were 66.9, 74, and 35.6 ppb, respectively. The average VOC OFP at the 20 gridded sites was 53.9 ppb, and the high VOC OFPs were concentrated at the XF, ZYT, LTC, CT, and HC sites (Fig. 11b). Of the sites, XF exhibited the highest OFP of 140.7 ppb, followed by ZYT and LTC with VOC OFPs of 87.9 and 75.7 ppb, respectively. The WQ site exhibited the lowest OFP of 21.9 ppb, followed by the XX and XHT sites with OFPs of 30.76 and 30.79 ppb, respectively. At the CB, DHS, and QL sites, alkenes and OVOCs played a dominant role in the OFP, accounting for 30.1 %–40.2 % and 31.1 %–39.4 %, respectively (Fig. 11b). For the gridded sampling sites, the overall OFP in Xi’an was significantly affected by alkenes and aromatics, accounting for 39.7 %–67.7 % and 7.2 %–23.2 %, respectively. The top 10 VOC OFP species at the CB, DHS, QL, and grid sampling sites accounted for 67.3 %, 66.8 %, 73.3 %, and 73.3 % of the total OFP, respectively (Table 2). Of these species, ethylene, acetaldehyde, isoprene, and m / p-xylene were the greatest contributors to Xi’an, reaching 44.2 %–54.5 %.

#### 3.4.2 VOC–NO\textsubscript{x}–O\textsubscript{3} sensitivity

The relationship between the ozone production rates (P (O\textsubscript{3})), anthropogenic VOC (AVOC) reactivity, and NO\textsubscript{x} reactivity of the CB, DHS, and QL sites during the observation period was shown in Fig. 12. The black curve in Fig. 12
represents the $P \left( \text{O}_3 \right)$ contour, and the black straight line represents the connection line of the $P \left( \text{O}_3 \right)$ turning point (ridgeline), whose slope represents the photochemical parameter $k_{\text{NO}_x}/k_{\text{VOCs}}$ (Jiang et al., 2018). When the site’s $k_{\text{NO}_x}/k_{\text{VOCs}}$ value is located above the ridgeline, it means that ozone formation is under VOC-limited regime; otherwise it means that ozone formation is under NO$_x$-limited regime. It can be seen from Fig. 12 that the ozone generation of the QL site is located in the NO$_x$-limited regime, and reducing NO$_x$ can effectively control ozone generation. The ozone generation of the DHS site is located in the VOC-limited regime, and reducing VOCs can effectively control ozone generation. However, the CB site is located in the transition regime between VOC- and NO$_x$-limited regimes. Therefore, simultaneous reduction of VOC and NO$_x$ concentration should be considered at the CB site to achieve the purpose of controlling O$_3$.

4 Conclusions

In this study, a campaign of field observations and VOC grid sampling was conducted in Xi’an from 20 June to 20 July 2019. During the observation period, the average VOC concentrations at the CB, DHS, QL, and gridded sampling sites were 27.8 ± 8.9, 33.8 ± 10.5, 15.5 ± 5.8, and 29.1 ± 8.4 ppb, respectively. Overall, the concentrations of VOCs in the Xi’an urban sites were approximately twice those of the rural sites. Due to a lower diffusion capacity and higher conversion capacity, the O$_3$ concentration in Xi’an urban areas (CB and DHS) often exceeded the national hourly standard of 200 µg/m$^3$ (approximately 101.9 ppb). In terms of the composition of VOCs at each site, alkanes and OVOCs were dominant in the VOC concentration, accounting for 25.7%–39.7% and 22.8%–47.4% of the TVOCs, respectively.

The PMF results demonstrated that vehicle exhaust, industry, combustion, paint solvent usage, and aging are the primary sources of VOCs in Xi’an. Of the sources, vehic-
Table 2. Top 10 VOC concentration, \(L_{OH}\), and OFP species in the CB, DHS, QL, and grid sampling sites.

<table>
<thead>
<tr>
<th>Site</th>
<th>Rank</th>
<th>Species</th>
<th>Concentration (ppb)</th>
<th>Concentration (L_{OH}) ((\times)10(^{-1})/s)</th>
<th>Species</th>
<th>OFP (ppb)</th>
</tr>
</thead>
<tbody>
<tr>
<td>CB</td>
<td>1</td>
<td>Acetone</td>
<td>4.9</td>
<td>13.8</td>
<td>Ethylene</td>
<td>11.8</td>
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<tr>
<td></td>
<td>2</td>
<td>Ethane</td>
<td>3.0</td>
<td>5.8</td>
<td>Acetaldehyde</td>
<td>10.2</td>
</tr>
<tr>
<td></td>
<td>3</td>
<td>Propane</td>
<td>1.9</td>
<td>2.7</td>
<td>Isoprene</td>
<td>6.0</td>
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<td></td>
<td>4</td>
<td>Acetaldehyde</td>
<td>1.6 (m/p)-Xylene</td>
<td>2.2</td>
<td>(m/p)-Xylene</td>
<td>3.8</td>
</tr>
<tr>
<td></td>
<td>5</td>
<td>Ethyne</td>
<td>1.4 (trans)-2-Butene</td>
<td>2.0</td>
<td>Methyl vinyl ketone</td>
<td>2.6</td>
</tr>
<tr>
<td></td>
<td>6</td>
<td>Ethylene</td>
<td>1.3 (1)-Butene</td>
<td>1.8</td>
<td>(1)-Butene</td>
<td>2.3</td>
</tr>
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<td>7</td>
<td>(n)-Butane</td>
<td>1.3 Propanal</td>
<td>1.4</td>
<td>Propene</td>
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<tr>
<td></td>
<td>8</td>
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<td>Ethylene</td>
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<td>0.8</td>
<td>(n)-Butane</td>
<td>1.5</td>
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</table>

cle exhaust was the primary source of VOC emissions in Xi’an, and the contributions of vehicle exhaust to the VOCs at the CB, DHS, and QL sites were 41.3 %, 30.6 %, and 23.6 %–41.4 %, respectively. In urban areas, industrial emissions were the second-largest source of VOC emissions, accounting for 29.7 % and 30.6 % at the CB and DHS sites, respectively. However, the contribution of aging sources was greater in rural areas, accounting for 19.2 % at the QL site.

Based on the 24 h backward trajectories and PSCF analysis for Xi’an, in the urban areas, the VOC concentration was significantly affected by the southeast trajectory, and the high PSCF values were primarily observed to the east and south of Xi’an. In rural areas, the VOC concentration was significantly affected by the northeast trajectory, and the high PSCF values were primarily observed to the east of Xi’an. The highest PSCF values appeared in the area near the CB, DHS, and QL sites, which indicates that Xi’an has a strong local source.

In Xi’an, alkenes, aromatics, and OVOCs played a dominant role in secondary transformation. Isoprene, acetalde-
Table 3. Air mass cluster analysis results for the CB, DHS, and QL sites.

<table>
<thead>
<tr>
<th>Site</th>
<th>Cluster</th>
<th>Air mass origins</th>
<th>Ratio (%)</th>
<th>TVOCs (ppb)</th>
<th>P_Ratio (%)</th>
<th>P_TVOCs (ppb)</th>
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</thead>
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<td>8.4</td>
<td>38.3</td>
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<tr>
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<tr>
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<td>36.4</td>
<td>15.3</td>
<td>34.3</td>
<td>23.3</td>
</tr>
</tbody>
</table>

LDT means long-distance trajectory. M-LDT means medium-long-distance trajectory. P_Ratio means the proportions of the pollution trajectories in each type of clustering trajectory. P_TVOCs means the average TVOC concentration of the pollution trajectories.

Figure 11. Contributions of different groups of VOCs to (a) LOH and (b) OFP in different sites in Xi’an.

Figure 12. The ozone production rate ($P(O_3)$) contour diagram versus anthropogenic VOCs (AVOCs) and NO$_x$ using the empirical kinetic modeling approach at CB, DHS, and QL sites.

hyde, ethylene, and 1-butene were the greatest contributors to LOH in Xi’an, reaching 36.4%–63.0%. Ethylene, acetaldehyde, isoprene, and m/p-xylene were the greatest contributors to OFP in Xi’an, reaching 44.2%–54.5%. These are the major species of concern for reducing O$_3$ pollution in Xi’an. The VOC–NO$_x$–O$_3$ sensitivity analysis results showed that the ozone generation of the DHS site is located in the VOC-limited regime, of CB in the transition regime between VOC- and NO$_x$-limited regimes, and of QL sites in the NO$_x$-limited regime. Therefore, reducing VOC concentration at the DHS site, reducing VOC and NO$_x$ concentration at CB site, and reducing NO$_x$ concentration at QL site can effectively control ozone generation.

Data availability. The underlying research data can be accessed upon contact with the corresponding author (Xin Li: li_xin@pku.edu.cn).

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Author contributions. XL and YZ designed the study. MS and XL analyzed the data and wrote the paper. XL, LZ, HD, KL, SC, JC, and NZ planned the locations of field observations and VOC grid sampling sites and built field observations sites. MS, SY, XY, YY, and SZ contributed to field observations and preprocessed data. KL, QC, and XY contributed to VOC grid sampling. MS and XL contributed to the revision of the paper. All authors contributed to the discussion and improvement of the paper.

Competing interests. The authors declare that they have no conflict of interest.

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References


Guenther, A. B., Zimmerman, P. R., Harley, P. C., Monson, R. K., and Fall, R.: Isoprene and monoterpene emission rate variability: A comparison to MEGAN2.1: an extended and updated framework for mod-


Liu, B., Liang, D., Yang, J., Dai, Q., Bi, X., Feng, Y., Yuan, J., Xiao, Z., Zhang, Y., and Xu, H.: Characterization and source apportionment of volatile organic compounds based on 1-year of

https://doi.org/10.5194/acp-21-4939-2021 Atmos. Chem. Phys., 21, 4939–4958, 2021


