Primary and secondary sources of ambient formaldehyde in the Yangtze River Delta based on Ozone Mapping and Profiler Suite (OMPS) observations

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Received: 13 November 2018 – Discussion started: 2 January 2019
Revised: 3 April 2019 – Accepted: 16 April 2019 – Published: 20 May 2019

Abstract. Formaldehyde (HCHO) in the ambient air not only causes cancer but is also an ideal indicator of volatile organic compounds (VOCs), which are major precursors of ozone (O₃) and secondary organic aerosol (SOA) near the surface. It is meaningful to differentiate between the direct emission and the secondary formation of HCHO for HCHO pollution control and sensitivity studies of O₃ production. However, understanding of the sources of HCHO is still poor in China, due to the scarcity of field measurements (both spatially and temporally). In this study, tropospheric HCHO vertical column densities (VCDs) in the Yangtze River Delta (YRD), East China, where HCHO pollution is serious, were retrieved from the Ozone Mapping and Profiler Suite (OMPS) onboard the Suomi National Polar-orbiting Partnership (Suomi-NPP) satellite from 2014 to 2017; these retrievals showed good agreement with the tropospheric HCHO columns measured using ground-based high-resolution Fourier transform infrared spectrometry (FTS) with a correlation coefficient (R) of 0.78. Based on these results, the cancer risk was estimated both nationwide and in the YRD region. It was calculated that at least 7840 people in the YRD region would develop cancer in their lives due to outdoor HCHO exposure, which comprised 23.4 % of total national cancer risk. Furthermore, the contributions of primary and secondary sources were apportioned, in addition to primary and secondary tracers from surface observations. Overall, the HCHO from secondary formation contributed most to ambient HCHO and can be regarded as the indicator of VOC reactivity in Hangzhou and in urban areas of Nanjing and Shanghai from 2015 to 2017, due to the strong correlation between total HCHO and secondary HCHO. At industrial sites in Nanjing, primary emissions more strongly influenced ambient HCHO concentrations in 2015 and showed an obvious decreasing trend. Seasonally, HCHO from secondary formation reached a maximum in summer and a minimum in winter. In the spring, summer, and autumn, secondary formation had a significant effect on the variation of ambient HCHO in urban regions of Nanjing, Hangzhou, and Shanghai, whereas in the winter the contribution from secondary formation became less significant. A more thorough understanding of the variation of the primary and secondary contributions of ambient HCHO is needed to develop a better knowledge regarding the role
of HCHO in atmospheric chemistry and to formulate effective control measures to decrease HCHO pollution and the associated cancer risk.

1 Introduction

In recent years, air pollution has become increasingly serious in China (Tang et al., 2012; Shi et al., 2014; Rohde and Muller, 2015), and it currently poses a grievous risk to human health (Wang and Mauzerall, 2006). Previous studies have indicated that air pollution has killed more people worldwide than tuberculosis, AIDS, breast cancer, and malaria (Yang et al., 2013; World Health Organization, 2014a, b). According to the US Environmental Protection Agency (EPA), HCHO has been identified as the major carcinogen among the 187 hazardous air pollutants (HAPs), and exposure to levels as low as 1 μg m⁻³ (about 0.7 ppb at standard temperature and pressure) of HCHO over a human lifetime could cause up to 13 in every million people to develop lung and nasopharyngeal cancer (Zhu et al., 2017a). HCHO not only directly affects public health but also plays a significant role in atmospheric photochemistry (Levy, 1971). The photolysis of HCHO leads to the production of the HO₂ radical (OH + HO₂) and subsequently affects the oxidative capacity of the atmosphere (Volkamer et al., 2010). Moreover, HCHO may contribute to O₃ pollution via photochemical reactions (Haagen-Smit, 1950; Carter, 1994; Russell et al., 1995), which favor the formation of secondary organic aerosol (SOA) by providing the OH radical (Yang et al., 2018). Therefore, monitoring and controlling the ambient HCHO concentration are of great significance for public health.

HCHO can be directly emitted into the atmosphere from biogenic sources, such as biomass burning and vegetation (Lee et al., 1997; Finlayson-Pitts and Pitts, 1999; Holzinger et al., 1999; Andreae and Merlet, 2001), and anthropogenic activities, such as vehicles emissions, industrial emissions, and coal combustion (Carlier et al., 1986; Williams et al., 1990; Hoekman, 1992; Carter, 1995; Anderson et al., 1996; Kean et al., 2001; Klimont et al., 2002; Reyes et al., 2006; Wei et al., 2008; Wang et al., 2013; Dong et al., 2014; Liu et al., 2017). The direct emission of HCHO is mainly due to incomplete combustion and is closely related to the emission of CO. Furthermore, HCHO can be formed from the atmospheric oxidation of volatile organic compounds (VOCs) (Althuiller, 1993; Carter, 1995; Seinfeld and Pandis, 2016) which also leads to the formation of O₃ (Levy, 1971; Crutzen, 1979; Warneke et al., 2004; Duan et al., 2008). In order to effectively control HCHO concentrations and improve air quality, it is necessary to determine the primary emission sources and secondary formation pathways of ambient HCHO.

In previous studies, the ratio of the Ozone Monitoring Instrument (OMI) tropospheric column of HCHO and NO₂ have been used as indicators of the sensitivity regime of O₃ formation (Duncan et al., 2010; Witte et al., 2011; Liu et al., 2016; Su et al., 2017), under the assumption that HCHO is a proxy for the reactivity of VOCs. Secondary HCHO is produced with the formation of peroxy radicals (RO₂), and is therefore positively correlated with RO₂; thus, it can be identified as a proxy for total VOC reactivity (Carter, 1994). Therefore, strictly speaking, only the HCHO concentration that stems from secondary formation can be used to analyze the sensitivity regime of O₃ production, combined with the NO₂ concentration. As secondary formation is the dominant source of HCHO on a global scale (De Smedt et al., 2008), total HCHO is usually adopted to replace secondary HCHO as the indicator of total VOC reactivity. However, in some regions that are subject to strong human activities, the contribution of primary emissions to ambient HCHO cannot be ignored (Ma et al., 2016). Hence, the separation of secondary HCHO from primary HCHO would be favorable to develop a better understanding of O₃ formation sensitivity, especially in urban areas.

Previous studies have attempted to identify contributions to ambient HCHO from direct emission and secondary formation via the use of linear multiple regression analysis. Because of the strong relation between primary HCHO and CO emission and the formation of HCHO and O₃ from the oxidation of VOCs, CO and O₃ can be used as the respective tracers for primary emission and secondary formation of HCHO. Several studies have been carried out on primary and secondary sources of HCHO using the CO–O₃ tracer pair. For instance, contributions from primary and secondary sources to ambient HCHO were characterized by statistical analogy to CO and O₃ in Houston, TX in the summer of 2000 (Friedfeld et al., 2002), in Beijing during the 2008 Olympic Games (Li et al., 2010), and in Hong Kong from winter of 2012 to autumn of 2013 (Lui et al., 2017). Hong et al. (2018) estimated contributions from primary emission and secondary formation to HCHO measured by MAX-DOAS using the CO–O₃ (O₃ + NO₂) tracer pair which was measured by the Sensor Networks for Air Quality (SNAQ) in the winter of 2015 in Yangtze River Delta. Garcia et al. (2006) indicated that using the CO–CHOCHO tracer pair to separate different sources of HCHO was more reasonable than using CO–O₃ because CHOCHO has a similar atmospheric lifetime to HCHO. However, continuous observations of CHOCHO are unfortunately not available.

Previous studies have often concentrated on contributions from different sources at one specific site and within a short time period; this is due to the constraints of available measurements of HCHO and its tracers with respect to primary emissions and secondary formation. In China, CO and O₃ measurements can be obtained from the China National Environmental Monitoring Center (CNEMC) network. However, HCHO is not measured at these stations. There are various analytical methods available to measure HCHO concentrations, including the following: Fourier transform in-
frared spectroscopy (FTIR; Lawson et al., 1990); differential optical absorption spectroscopy (DOAS; Platt and Perner, 1980); tunable diode laser absorption spectroscopy (TD-LAS); the 2,4-dinitrophenylhydrazine (DNPH) method, in which HCHO is captured using DNPH and then analyzed by high-performance liquid chromatography (HPLC; Fung and Grosjean, 1981); and proton transfer reaction mass spectrometry (PTR-MS; Karl et al., 2003). Nevertheless, it is difficult to perform long-term observations of the HCHO concentration over a wide area using ground-based measurements. However, long-term information regarding ambient HCHO on a global scale can be accessibly obtained from space-based remote sensors such as Global Ozone Monitoring Experiment (GOME; Martin et al., 2004), the Scanning Imaging Absorption Spectrometer for Atmospheric CHartography (SCIAMACHY; Bovensmann et al., 1999), the GOME-2 instruments (De Smedt et al., 2012, 2015), the Ozone Monitoring Instrument (OMI; González Abad et al., 2015), the Ozone Mapping and Profiler Suite (OMPS; C. Li et al., 2015; González Abad et al., 2016) and the TROPOspheric Monitoring Instrument (TROPOMI; De Smedt et al., 2018). However, HCHO observations from SCIAMACHY and TROPOMI are only available until 2011 (Shah et al., 2018) and after October 2017, respectively; the GOME-2A instrument suffers degradation issues (De Smedt et al., 2012); OMI observations are easily affected by the instrumental “row anomaly” (González Abad et al., 2015); and the spatial (80 km × 40 km) and temporal (within 1.5 d) resolution of GOME-2B is even lower than OMPS (Munro et al., 2016). After considering the long-term record and data quality, HCHO observations from OMPS were adopted in this study. Due to the low global coverage of the Tropospheric Emission Spectrometer (Aura-TES; Luo et al., 2002, 2007) and the increasing noise of the Atmospheric InfraRed Sounder (Aqua-AIRS) after 2004 (Pagano et al., 2012), which both measure tropospheric CO in close temporal proximity to the OMPS observations, obtaining a precise CO concentration in the YRD after 2013 from satellite data was difficult. Therefore, CO and O₃ concentrations at the surface, simultaneously monitored by the China National Environmental Monitoring Center (CNEMC) network, were used in this study.

In this paper, we focus on contributions from primary and secondary sources to ambient HCHO from 2015 to 2017 in Nanjing (the capital of Jiangsu Province), Hangzhou (the capital of Zhejiang Province) and Shanghai; these are all megacities in the YRD region and have populations of more than 8 million people. The YRD region is located on the al-luval plains where the Yangtze River drains into the East China Sea, and consists of Jiangsu and Zhejiang provinces and the Shanghai municipality. The YRD region is one of the most important economic areas in China and is also a rapidly developing industrial sector. Along with the very rapid development of economy and industry, the YRD region is currently suffering from serious air pollution (H. Li et al., 2015; Gao et al., 2016; T. Wang et al., 2017; Wang et al., 2018). Details regarding the measurement of specific air pollutants are described in Sect. 2, whereas the spatiotemporal distribution of HCHO VCDs in the YRD region, as well as the contributions from primary and secondary sources to ambient HCHO are shown in Sect. 3.

2 Measurements and methodology

2.1 OMPS HCHO observation

The Ozone Mapping and Profiling Suite Nadir Mapper (OMPS-NM), which is one of the OMPS suite of instruments, was launched on 28 October 2011 onboard the Suomi National Polar-orbiting Partnership (Suomi-NPP) satellite. The Suomi-NPP is a polar sun-synchronous satellite with an average altitude of 824 km (Flynn et al., 2014). It crosses the Equator each afternoon at about 13:30 LT (local time) on the ascending node. The OMPS-NM combines a single grating and a 340 × 740 pixel charge-coupled device (CCD) detector to measure UV radiation every 0.42 nm from 300 to 380 nm with a 1.0 nm full width at half maximum (FWHM) resolution. A recent analysis showed that the OMPS-NM sensor has a signal-to-noise ratio of 2000:1 or better within the wavelength range of 320 to 370 nm (Seftor et al., 2014). It has a 110° cross-track field of view (FOV) providing high temporal (daily global coverage) resolution, and measurements are combined into 35 cross-track bins giving a spatial resolution of 50 km × 50 km in OMPS-NM standard Earth science mode. The OMPS-NP Nadir Mapper Earth View Level 1B data, available at https://search.earthdata.nasa.gov (last access: 8 May 2019), were downloaded to retrieve the HCHO slant column density (SCD). Details regarding the fitting settings for the HCHO SCD retrieval are given in González Abad et al. (2016). In this study, the vertical profiles of HCHO from WRF-Chem modeling were used to calculate the air-mass factor (AMF). The model configuration was described in detail in our previous study (Su et al., 2017). Finally, the HCHO VCD was calculated using the AMF as follows:

\[
VCD = \frac{SCD}{AMF}
\] (1)

2.2 FTS tropospheric HCHO column measurement

The ground-based FTS instrument, located west of Hefei (117.17°E, 31.9°N), is a candidate station for the Network for Detection of Atmospheric Composition Change (NDACC, http://www.ndacc.org/, last access: 8 May 2019). The global NDACC can perform simultaneous retrieval of mixing ratio profiles for trace gases, such as O₃, CO, CH₄, NO₂, and HCHO, using FTS (Kurylo, 1991; Notholt et al., 1995; Vigouroux et al., 2009), and the retrieval results have been widely used for atmospheric chemistry research and the validation of satellite observations (Yuan et al., 2015; Sun et
The observation system consists of a high-resolution Fourier transform spectrometer (IFS 125HR), a solar tracker (Tracker-A solar 547), and a weather station (ZEN0-3200). The FTS solar spectra measurements were performed over a broad spectral range from 600 to 4500 cm$^{-1}$ with a spectral resolution of 0.005 cm$^{-1}$, and the HCHO spectra were recorded in the range from 2400 to 3310 cm$^{-1}$. The vertical profile of HCHO was retrieved using the SFIT4 algorithm (version 0.9.4.4). Detailed retrieval settings for HCHO are listed in Sun et al. (2018). Daily a priori profiles of pressure, temperature, and H$_2$O are obtained from the National Centers for Environmental Protection and the National Center for Atmospheric Research (NCEP/NCAR) reanalysis. And a priori profiles of HCHO and its interfering gases were taken from a dedicated Whole Atmosphere Community Climate Model (WACCM). For HCHO observation, the four microwindows (MW) which are centered at around 2770 cm$^{-1}$ were selected for satellite validation, and a de-weighting signal-to-noise ratio (SNR) of 600 was used. The instrument line shape (ILS) is described by analyzing HBr cell spectra using LINEFIT14.5 software. Finally the tropospheric HCHO partial column was calculated by integrating the retrieved profiles with the air-mass profile within 15 km of the retrievals as follows:

$$P_{C_{trop}} = \int_{0}^{15} X_t \times A_m, \quad (2)$$

where $X_t$ is the retrieved profile of HCHO, $A_m$ is the air-mass profile, and $P_{C_{trop}}$ is the tropospheric HCHO column. The FTS total systematic error of the HCHO retrieval is less than 3.3 %, and the total random error is less than 9.6 % (Sun et al., 2018). Due to the high accuracy, the trace gas concentrations measured using FTS have been widely used to validate corresponding satellite products (Yamamori et al., 2006; Jones et al., 2009; Yurganov et al., 2010; Reuter et al., 2011). The FTS total systematic error of the HCHO retrieval is less than 3.3 %, and the total random error is less than 9.6 % (Sun et al., 2018). Due to the high accuracy, the trace gas concentrations measured using FTS have been widely used to validate corresponding satellite products (Yamamori et al., 2006; Jones et al., 2009; Yurganov et al., 2010; Reuter et al., 2011).

2.3 The China National Environmental Monitoring Center (CNEMC) network

Surface air pollutants monitored on an hourly basis by CNEMC were provided by the Ministry of Environment Protection of the People’s Republic of China (http://106.37.208.233:20035/, last access: 8 May 2019). The national air quality monitoring network consisted of about 950 sites in 2013, and had extended to 1597 sites in 454 major cities by 2017. Three trace gases, including CO, O$_3$, and SO$_2$, fine particles (PM$_{10}$), and PM$_{2.5}$ were simultaneously measured by all sites, and the data have been widely used in various research (W.-N. Wang et al., 2017; Li et al., 2018; Liu et al., 2018; X. Lu et al., 2018). In this study, CO and O$_3$ data from 2104 to 2017 were used. Data quality controls similar to those in previous studies (Barrero et al., 2015; He et al., 2017; Li et al., 2018) were applied to remove data outliers. Briefly, all hourly data at a specific monitoring site were transformed into $z$ scores, and then the transformed data ($Z_i$) were removed if they met one of the following conditions: (1) the absolute $Z_i$ was larger than 4 ($|Z_i| > 4$), (2) the increment of $Z_i$ from the previous hourly value was larger than 9 ($|Z_i - Z_{i-1}| > 9$), or (3) the ratio of the $z$ score to its centered moving average of order 3 (MA3) was larger than 2 ($Z_{i-1} + Z_i + Z_{i+1} > 2$).

3 Results

3.1 Comparison with FTS tropospheric HCHO

In order to validate the OMPS observations using FTS measurements, the HCHO tropospheric column measured using FTS was averaged around the OMPS satellite overpass time. The OMPS tropospheric HCHO column was selected for satellite pixels within 50 km of the Hefei site and with cloud fraction less than 20 %, and these pixels were averaged to minimize the random noise (Y. Wang et al., 2017). Figure 1a shows that tropospheric HCHO VCDs observed by OMPS and FTS are in good agreement ($R = 0.78$). The coincident time series of both data indicate that OMPS HCHO observations successfully capture the maximum HCHO concentration in summer and the minimum HCHO concentration in winter (Fig. 1b). However, it is important to note that the OMPS HCHO observations underestimate the tropospheric HCHO column by 0 %–60 % compared with the FTS results, especially in summer; these results are consistent with a previous study by De Smedt et al. (2015). The underestimation from OMPS was attributed to the spatial average from satellite observations, errors from HCHO SCDs affected by spectral fitting, and AORF calculation error which were affected by scattering weights calculated by the radiative transfer model and HCHO vertical profiles modeled by WRF-Chem (Zhu et al., 2016). Most of the atmosphere HCHO was concentrated in the troposphere. The monthly averaged vertical profiles of HCHO shown in Fig. 2a indicate that the structure of the HCHO vertical profiles in different months was consistent. The HCHO concentration decreased by 70 % with an increase in the height from 0.7 to 2 km and continued to decrease slowly in the troposphere above 2 km. The ratio of the HCHO column below 1 km in the tropospheric HCHO column remained at ~ 67 % and did not show seasonal variation (Fig. 2b). The HCHO column below 1 km measured by FTS and the OMPS tropospheric HCHO column also showed good agreement ($R = 0.78$, Fig. S1 in the Supplement). The slope of 0.83 for the linear regression analysis between the FTS HCHO column below 1 km and the OMPS tropospheric column was higher than the slope of 0.51 for the regression between OMPS data and the tropospheric HCHO VCD.
Figure 1. Correlation analysis (a) and time series (b) of tropospheric HCHO VCDs measured by OMPS and FTS from 1 July 2015 to 20 November 2017.

Figure 2. (a) The monthly averaged HCHO vertical profiles in the troposphere measured by FTS. (b) The monthly averaged tropospheric HCHO column (black line), the HCHO column below 1 km (red line) measured by FTS, and the ratio of the HCHO column below 1 km in the tropospheric HCHO column (blue line).

from FTS, indicating that the OMPS tropospheric HCHO column better represents the HCHO column below 1 km. Furthermore, the mean of the relative differences between the HCHO column below 1 km measured by FTS and the tropospheric HCHO column measured by OMPS is zero. In previous studies, a uniform and constant correction factor has been applied to account for the bias between the tropospheric HCHO column observed by satellite and that measured by aircraft (e.g., Anderson et al., 2016). In this study, we regard the tropospheric HCHO column observed by OMPS as that below 1 km measured by FTS in terms of their numerical value. Assuming that HCHO mixes well within the 1 km above the surface, the mixing ratio of HCHO can be calculated as follows:

\[ M(\text{ppbv}) = \frac{1.25 \times \text{VCD (molecules cm}^{-2}\text{)}}{B \times \Delta p(\text{atm})}, \]  

where \( M \) is the mixing ratio of HCHO, \( B \) is unit conversion factor of DU to molecules cm\(^{-2}\) which is \(2.688 \times 10^{16}\), VCD is the OMPS tropospheric HCHO VCDs, and \( \Delta p \) is the pressure difference between the surface and 1 km (Ziemke et al., 2001; Lee et al., 2008).

3.2 Spatiotemporal distribution of HCHO VCDs in the Yangtze River Delta

The average HCHO VCD over China in August 2017 was \(6.50 \times 10^{15}\) molecules cm\(^{-2}\), which would lead up to 50 people in every million to develop cancer. The total national cancer risk was estimated by combining the averaged HCHO VCDs with the population, and it was estimated that at least 33,500 people in China would develop lung and nasopharyngeal cancer in their lifetime due to outdoor HCHO exposure using Eq. (4):
\[ T = \sum_i P_i \times C_i \]  
\[ Y = \frac{T}{L}, \]  
where \( T \) is the total number of people in China who may develop cancer due to outdoor HCHO exposure, \( P_i \) and \( C_i \) represent the respective population and average cancer risk of provincial level administrative region \( i \), \( Y \) is the number of cancer cases per year, and \( L \) is the life expectancy. Considering the life expectancy of 76.4 years (2015) in China (http://data.stats.gov.cn/easyquery.htm?cn=C01&zb=A0304&sj=2016, last access: 8 May 2019, in Chinese), 439 cancer cases per year in China are caused by high outdoor HCHO concentrations (Eq. 5). The HCHO VCDs in most of western China, e.g., Yunnan Province, Tibet, Gansu Province, Qinghai Province, the Xinjiang Uygur Autonomous Region, Ningxia Province, and the Inner Mongolia Autonomous Region, were lower than the national average (Fig. 3). The HCHO VCDs in central China were higher than the national average, and the distribution of HCHO VCDs was homogenous. The highest HCHO VCDs occurred in eastern China, specifically in the Beijing–Tianjin–Hebei region, the northern Henan Province, the western Shandong Province, the YRD region, and the Pearl River Delta (PRD) region, with the most severe HCHO pollution occurring in the YRD. The high outdoor HCHO concentration in the YRD increases the cancer risk for inhabitants, with up to 155 people in every million suffering cancer, which is about four times the cancer risk reported in the US (Zhu et al., 2017a). The total cancer risk in the YRD comprises 23.4 % of total national cancer risk in China, and about 7840 people in the YRD may develop cancer during their lifetime due to severe HCHO pollution. Therefore, effectively controlling the ambient HCHO concentration in the YRD is an urgent issue.

In addition, the YRD has been identified as the region with the highest emissions of VOCs (Qiu et al., 2014; Wu et al., 2015); therefore, the determination of the contribution from primary sources and secondary formation to ambient HCHO is needed to formulate appropriate control measures.

In the YRD region, the highest HCHO VCDs occurred in southwestern Jiangsu Province (e.g., Nanjing, Changzhou, Wuxi, and Suzhou cities), northern Zhejiang Province (e.g., Hangzhou, Jiaxing, Huzhou, and Ningbo cities), and Shanghai in 2015 (Fig. 4a). The spatial distribution of annual mean HCHO VCDs in 2016 and 2017 was similar to that in 2015. The HCHO VCDs in the YRD region showed a fluctuating trend from 2015 to 2017 (Fig. 4b, c). The HCHO concentration in Jiangsu Province rose from 2015 to 2016, whereas only HCHO VCDs in southwestern Jiangsu Province showed...
Figure 4. (a) The spatial distribution of annual mean tropospheric HCHO column observed by OMPS in the YRD in 2015, and (b) changes in the HCHO VCDs between 2016 and 2015 and (c) between 2017 and 2016. A decreasing trend from 2016 to 2017. The HCHO concentration in Shanghai and Hangzhou showed a similar trend: they clearly decreased from 2015 to 2016 and then increased from 2016 to 2017. An opposite trend was observed for HCHO VCDs in other cities in Zhejiang Province.

3.3 Determination of primary and secondary contributions to ambient HCHO

The statistical analysis of simultaneous real-time measurements of HCHO, CO, and O$_3$ can be described using a multiple regression model:

$$[C_{\text{HCHO}}] = \beta_0 + \beta_1 \times [C_{\text{CO}}] + \beta_2 \times [C_{\text{O}_3}],$$  \hspace{1cm} (6)

where $\beta_0$, $\beta_1$, and $\beta_2$ are the coefficients fitted by the model (Garcia et al., 2006), and $[C_{\text{HCHO}}]$, $[C_{\text{CO}}]$, and $[C_{\text{O}_3}]$ represent the concentrations or the transformations of concentrations for HCHO, CO, and O$_3$, respectively. Similar to Garcia et al. (2006), we applied 10 different transformations on the time series of HCHO, CO, and O$_3$: linear (no transformation), natural log, square root, second power, third power, inverse, inverse of natural log, inverse of square root, inverse of second power, and inverse of third power. Linear regression without transformation was identified as the best model (Table S1). The relative contributions of background concentration, primary emissions, and secondary formation to the ambient HCHO can be calculated using the following equations:

$$R_{\text{Primary}} = \frac{\beta_1 \times [C_{\text{CO}}]}{\beta_0 + \beta_1 \times [C_{\text{CO}}] + \beta_2 \times [C_{\text{O}_3}]} \times 100\%,$$ \hspace{1cm} (7)

$$R_{\text{Secondary}} = \frac{\beta_2 \times [C_{\text{O}_3}]}{\beta_0 + \beta_1 \times [C_{\text{CO}}] + \beta_2 \times [C_{\text{O}_3}]} \times 100\%,$$ \hspace{1cm} (8)

where $R_{\text{Primary}}$ represents the contribution to ambient HCHO from primary sources (e.g., industrial and vehicle emissions), $R_{\text{Secondary}}$ represents the contribution to ambient HCHO from secondary sources (i.e., photochemical VOC oxidation), and $R_{\text{Background}}$ represents background contributions to the ambient HCHO which cannot be classified as primary or secondary contributions. In previous studies in the YRD (Wang et al., 2015; Ma et al., 2016), the background HCHO concentration of 1 ppbv has been selected to represent the HCHO concentration under regional conditions. Therefore, the $\beta_0$ coefficient is fixed at 1 ppbv in the multiple linear regression in this study.

In a previous study, Ma et al. (2016) analyzed the primary emissions and secondary formation of HCHO from 15 April to 1 May 2015 at an industrial zone in Nanjing, which is located less than 15 km from one of our research sites, Maigaoqiao (MGQ) station, using the in situ measurement of primary tracers (i.e., benzene, toluene, and CO) and a secondary tracer (i.e., O$_3$). Ma et al. (2016) found that the average relative contributions of industry-related emissions, secondary formation, and background were 59.2 %, 13.8 %, and 27 %, respectively. In this study, the average relative contributions from primary sources, secondary formation, and background to HCHO at MGQ station during the same period were 50.5 %, 30.8 %, and 18.7 %, respectively. We note that the relative contribution from primary emissions was similar to that reported by in Ma et al. (2016), with a difference of 8.7 %. However, the relative contribution from secondary formation was 17 % larger and the relative contribution from background was 8.3 % smaller than the values reported...
ported in Ma et al. (2016). It should be noted that the contribution of secondary formation we calculate is the maximum daily value, as the overpass time of the OMPS is 13:30 LT every day. Therefore, it is reasonable to assume that the contribution of secondary formation we calculate is also larger due to the favorable conditions for photochemical reactions at noon (Hong et al., 2018). In conclusion, the tropospheric HCHO column observed by OMPS can be used to analyze primary and secondary contributions to ambient HCHO via a multiple linear regression fit, and CO and O\(_3\) concentrations measured by CNEMC Network can be used as indicators of primary emission sources and secondary formation sources, respectively.

### 3.3.1 Primary and secondary sources of ambient HCHO in Nanjing

The distribution of the three stations selected to explore the primary and secondary sources of HCHO is shown in Fig. 5a, and the heavy industrial zones are delineated using red lines (Zheng et al., 2015). In order to ensure the representativeness of the regression model, only data fulfilling the criteria in the analysis (correlation coefficient (\(R\)) larger than 0.6 and significance value lower than 0.05) were kept; the same data quality control was carried out for the Hangzhou and Shanghai analyses. We selected the MGQ site, located southwest of the industrial zones, the Ruijin Road site (RJR), located in the center of Nanjing, and the Xianlin University Town site (XLUT), located southeast the of industrial zones, to analyze the HCHO concentration from different sources in Nanjing. As presented in Table 1, the annual average HCHO concentrations at industrial sites were larger than those at an urban site in 2015 and 2017. The seasonal average HCHO concentration adheres to the following seasonal order in both the industrial zones and center of Nanjing (Fig. 6 and Table 3): summer > autumn > spring > winter.

On average, secondary formation was the dominant source of ambient HCHO at all the three sites in Nanjing (Table 1). However, primary emission was found to be the most significant source in winter at the RJR site (Fig. 6a, b), in winter of 2015 at the MGQ site (Fig. 6c, d), and from August to December in 2015 at the XLUT site (Fig. 6e, f). At the industrial sites, MGQ and XLUT, the annual average HCHO concentration from primary emissions decreased by 64 % and 63.44 % from 2015 to 2017 in comparison with 2015, respectively. As coal combustion is an important primary source of ambient HCHO (Wang et al., 2013; Liu et al., 2017), energy saving methods and the reduction of energy consumption, especially the reduction of coal consumption, caused the decrease of primary HCHO emission, according to the Nanjing municipal government (http://www.nanjing.gov.cn/xxgk/szf/201403/t20140321_2544036.html, last access: 15 August 2018, in Chinese). The variation of the HCHO concentration from different sources at MGQ was similar to that at XLUT due to the short distance between the two sites. However, the annual average HCHO concentration from primary emissions at XLUT was larger than that at MGQ. Primary emissions of HCHO at the urban site were smaller than those at industrial sites in 2015 and larger than those at suburban sites in 2016 and 2017. The correlation coefficients between primary HCHO and total HCHO or secondary HCHO and total HCHO can be used to compare the contribution from the two sources to the variation of HCHO: a larger correlation coefficient means a higher contribution to the variation of HCHO. At the urban site, secondary formation contributed more to the variation of HCHO concentration from 2015 to 2017 (\(R = 0.76\), Fig. 7a) than primary emission (\(R = 0.35\),...
Fig. 6. The time series of absolute (a, c, e) and relative (b, d, f) contributions of primary sources, secondary sources, and background sources to the HCHO concentration from December 2014 to November 2017 at the RJR site (a, b), the MGQ site (c, d), and the XLUT site (e, f) in Nanjing. Black dashed lines represent the threshold value for HCHO pollution.

Fig. S2a). While at the industrial sites, the contribution from secondary formation to the variation of the HCHO concentration was less evident ($R \leq 0.67$, Fig. 7b, c). At the XLUT site, primary HCHO emission was more significant to the variation of the HCHO concentration than secondary formation in 2015 (Table 2). With respect to the seasonal average, secondary formation contributed most to the ambient HCHO concentration, and reached a maximum in summer, due to the improvement of photochemical reaction (Wang et al., 2016), and a minimum in winter (Table 3). In spring, summer, and autumn, the variation of the ambient HCHO was affected more by secondary formation, whereas in winter, it was influenced more by primary emission in the urban region (Table 4).

### 3.3.2 Primary and secondary sources of ambient HCHO in Hangzhou

Three sites, including Xiasha (XS), Chengxiang Town (CXT), and the “Fifth District of Chaohui” (FDCH) which are located in the northeastern suburbs, the eastern suburbs, and the center of Hangzhou, respectively, were selected to explore the sources of HCHO in Hangzhou, as shown in Fig. 5b. There was no significant difference between the HCHO concentrations at the suburban sites (Table 1). The average HCHO concentration in 2016 was larger than the average concentrations in 2015 and 2017 at all three sites. As shown in Table 1, secondary formation, with respect to the annual average, contributed more to ambient HCHO than primary emissions from 2015 to 2017 in Hangzhou. It is interesting that primary emissions of HCHO showed a small increasing trend from 2015 to 2017 at suburban sites (Fig. 8a, c, Table 1). According to the environment bulletin released by the Hangzhou government, this may have been caused by the significant increase in the number of vehicles and the less effective control measures regarding ambient HCHO. Moreover, secondary HCHO also showed an important growing tendency from 2015 to 2016 and then began to decrease in...
Table 1. Annual average of the measured HCHO concentrations (ppbv) and absolute and relative contributions from different sources in Nanjing, Hangzhou, and Shanghai. The values in parentheses represent the contributions from different sources; lines 1–2 in each year refer to primary HCHO, lines 3–4 refer to secondary HCHO, and line 5 refers to measured HCHO.

<table>
<thead>
<tr>
<th>Year</th>
<th>Nanjing</th>
<th>Hangzhou</th>
<th>Shanghai</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>MGQ</td>
<td>RJR</td>
<td>XLUT</td>
</tr>
<tr>
<td></td>
<td>(industrial zone)</td>
<td>(urban area)</td>
<td>(industrial zone)</td>
</tr>
<tr>
<td>2015</td>
<td>Primary</td>
<td>2.528</td>
<td>1.288</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(36.86%)</td>
<td>(18.16%)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(48.13%)</td>
<td>(46.41%)</td>
</tr>
<tr>
<td>2016</td>
<td>Primary</td>
<td>0.933</td>
<td>1.411</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(17.75%)</td>
<td>(11.49%)</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(63.88%)</td>
<td>(59.26%)</td>
</tr>
</tbody>
</table>

Table 2. The Pearson correlation coefficient (R) between the measured HCHO and primary HCHO emissions and secondary HCHO formation in Nanjing, Hangzhou, and Shanghai from 2015 to 2017. Line 1 in each year refers to primary HCHO and line 2 refers to secondary HCHO.

<table>
<thead>
<tr>
<th>Year</th>
<th>Nanjing</th>
<th>Hangzhou</th>
<th>Shanghai</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>MGQ</td>
<td>RJR</td>
<td>XLUT</td>
</tr>
<tr>
<td></td>
<td>(industrial zone)</td>
<td>(urban area)</td>
<td>(industrial zone)</td>
</tr>
<tr>
<td>2015</td>
<td>Primary</td>
<td>0.14</td>
<td>0.21</td>
</tr>
<tr>
<td></td>
<td>Secondary</td>
<td>0.58</td>
<td>0.75</td>
</tr>
<tr>
<td>2016</td>
<td>Primary</td>
<td>−0.43</td>
<td>0.27</td>
</tr>
<tr>
<td></td>
<td>Secondary</td>
<td>0.82</td>
<td>0.71</td>
</tr>
<tr>
<td>2017</td>
<td>Primary</td>
<td>0.07</td>
<td>0.5</td>
</tr>
<tr>
<td></td>
<td>Secondary</td>
<td>0.7</td>
<td>0.77</td>
</tr>
</tbody>
</table>

2017. The variation of the HCHO concentration was mainly affected by secondary formation from 2015 to 2017, due to the strong correlation ($R > 0.7$; Fig. 7d, e, f). In all seasons, the seasonal average of secondary formation from 2015 to 2017 was larger than primary emissions in Hangzhou (Table 3). In contrast, primary emission were seen to exceed secondary formation and become the main source on individual days in winter of 2016 and in summer of 2017 at the XS site (Fig. 8a, b), in winter of 2016 at the CXT site (Fig. 8c, d), and in winter of 2016 and 2017 at urban site (Fig. 8e, f). In winter, the impact of primary emission on the variation of HCHO was more significant than secondary formation, whereas in spring, summer, and autumn, secondary formation became more important to the variation of ambient HCHO, except in the northeastern suburbs in autumn (Table 4).

3.3.3 Primary and secondary sources of ambient HCHO in Shanghai

The “Pudong New Area” site (PDNA), the Hongkou site (HK), and the “Dianshan Lake in Qingpu” site (DSL) were selected to represent east of the city center, north of the city center, and the suburbs, respectively (Fig. 5c). The annual average HCHO concentration decreased by 0.342 ppbv from 2015 to 2016 and increased by 0.529 ppbv from 2016 to 2017 at the HK site (Table 1). The annual average HCHO concentration at the PDNA site increased by 0.15 ppbv in 2016 and
Figure 7. Correlation analysis of total HCHO and secondary HCHO from 2015 to 2017 at the RJR site (a), the MGQ site (b), and the XLUT site (c) in Nanjing; at the XS site (d), the CXT site (e), and the FDCH site (f) in Hangzhou; and at the HK site (g), the PDNA site (h), and the DSL site (i) in Shanghai (significance values are equal to zero).

decreased by 0.056 ppbv in 2017, compared with that from the previous year. At the DSL site, the HCHO concentration was less than that at HK, except in 2016. The HCHO concentration at DSL initially increased by 0.179 ppbv from 2015 to 2016 and then decreased by 0.143 ppbv from 2016 to 2017. We can conclude that the ambient HCHO concentration has not been strictly controlled in recent years. With respect to the seasonal average, the HCHO concentration reached a maximum in summer and a minimum in winter in Shanghai, which concurs with findings from Nanjing and Hangzhou (Table 3). Moreover, the HCHO concentration in Shanghai in summer was larger than that in Nanjing and Hangzhou.

On average, secondary formation was the most significant contribution to ambient HCHO in Shanghai, except at the PDNA site in 2016 (Table 1 and Fig. 9). The contribution from secondary formation developed, showing a rising trend, and primary HCHO emission showed a decreasing trend from 2015 to 2017, at the suburban site (Fig. 9e, f, Table 1). The variation of the HCHO concentration was more strongly affected by secondary formation ($R > 0.70$, Fig. 7g, h, i) than primary emissions ($R < 0.43$, Fig. S2g, h, i) from 2015 to 2017. In summer and autumn, secondary formation was the largest source of ambient HCHO (Fig. 9a, b, c, d) and was the main factor influencing the variation of the HCHO concentration in urban regions of Shanghai. At PDNA site, in comparison, primary emissions ($R = 0.61$, Table 4) had a more significant influence on the variation of ambient HCHO than secondary formation ($R = −0.42$, Table 4) in winter, and it became the most important source of ambient HCHO in spring and winter of 2016. The secondary sources of HCHO in the three megacities showed a similar seasonal variation, and showed a similar order of influence...
Table 3. Seasonal average of the measured HCHO concentrations (ppbv) and absolute and relative contributions from different sources from 2015 to 2017 in Nanjing, Hangzhou, and Shanghai. The values in parentheses represent the contributions from different sources; lines 1–2 in each season refer to primary HCHO, lines 3–4 refer to secondary HCHO, and line 5 refers to measured HCHO.

<table>
<thead>
<tr>
<th>Season</th>
<th>Nanjing</th>
<th>Hangzhou</th>
<th>Shanghai</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>MGQ (industrial zone)</td>
<td>RJR (urban area)</td>
<td>XLUT (industrial zone)</td>
</tr>
<tr>
<td>Winter</td>
<td>Primary</td>
<td>1.398 (29.45%)</td>
<td>1.151 (32.01%)</td>
</tr>
<tr>
<td></td>
<td>Secondary</td>
<td>2.011 (47.15%)</td>
<td>1.469 (39.54%)</td>
</tr>
<tr>
<td></td>
<td>Measured</td>
<td>4.403</td>
<td>3.394</td>
</tr>
<tr>
<td>Spring</td>
<td>Primary</td>
<td>1.51 (27.04%)</td>
<td>0.981 (20.00%)</td>
</tr>
<tr>
<td></td>
<td>Secondary</td>
<td>3.017 (54.37%)</td>
<td>2.996 (59.04%)</td>
</tr>
<tr>
<td>Summer</td>
<td>Primary</td>
<td>1.047 (13.89%)</td>
<td>1.414 (20.64%)</td>
</tr>
<tr>
<td></td>
<td>Secondary</td>
<td>5.503 (54.37%)</td>
<td>4.484 (64.54%)</td>
</tr>
<tr>
<td>Autumn</td>
<td>Primary</td>
<td>1.103 (18.36%)</td>
<td>1.537 (26.40%)</td>
</tr>
<tr>
<td></td>
<td>Secondary</td>
<td>3.905 (63.95%)</td>
<td>3.462 (64.54%)</td>
</tr>
<tr>
<td></td>
<td>Measured</td>
<td>5.851</td>
<td>5.463</td>
</tr>
</tbody>
</table>

Table 4. The Pearson correlation coefficient ($R$) between the measured HCHO and primary HCHO emission and secondary HCHO formation in Nanjing, Hangzhou, and Shanghai in winter, spring, summer, and autumn. Line 1 in each season refers to primary HCHO and line 2 refers to secondary HCHO.

<table>
<thead>
<tr>
<th>Season</th>
<th>Nanjing</th>
<th>Hangzhou</th>
<th>Shanghai</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>MGQ (industrial zone)</td>
<td>RJR (urban area)</td>
<td>XLUT (industrial zone)</td>
</tr>
<tr>
<td>Winter</td>
<td>Primary</td>
<td>0.67 (0.61)</td>
<td>0.24 (0.61)</td>
</tr>
<tr>
<td></td>
<td>Secondary</td>
<td>−0.41 (−0.42)</td>
<td>−0.23 (−0.42)</td>
</tr>
<tr>
<td>Spring</td>
<td>Primary</td>
<td>0.11 (0.12)</td>
<td>0.21 (0.16)</td>
</tr>
<tr>
<td></td>
<td>Secondary</td>
<td>0.25 (0.56)</td>
<td>0.6 (0.39)</td>
</tr>
<tr>
<td>Summer</td>
<td>Primary</td>
<td>0.17 (0.1)</td>
<td>0.54 (0.22)</td>
</tr>
<tr>
<td></td>
<td>Secondary</td>
<td>0.45 (0.6)</td>
<td>0.59 (0.35)</td>
</tr>
<tr>
<td>Autumn</td>
<td>Primary</td>
<td>0.42 (0.26)</td>
<td>0.26 (0.31)</td>
</tr>
<tr>
<td></td>
<td>Secondary</td>
<td>0.71 (0.41)</td>
<td>0.68 (0.48)</td>
</tr>
</tbody>
</table>

4 Discussion

4.1 Reconsidering the proxy for VOC reactivity using the HCHO concentration

Duncan et al. (2010) determined the indicator of VOC and NO$_x$ controls on surface O$_3$ formation using HCHO VCDs.
Figure 8. The time series of absolute (a, c, e) and relative (b, d, f) contributions of primary sources, secondary sources, and background sources to the HCHO concentration from December 2014 to November 2017 at the XS (a, b), CXT (c, d), and FDCH (e, f) sites in Hangzhou. Black dashed lines represent the threshold value for HCHO pollution.

observed by OMI as the proxy for VOC reactivity in August between 2005 and 2007 in the US. This is reasonable, as the ambient HCHO concentration mainly stemmed from the atmospheric oxidation of isoprene in summer over North America (Palmer et al., 2003; Zhu et al., 2017b; Marvin et al., 2017); furthermore, the HCHO concentration from secondary formation contributed 87% during daytime in Los Angeles in the 1980s (Kawamura et al., 2000), and over 80% of the ambient HCHO concentration is from secondary formation over the eastern US (Luecken et al., 2006). However, the contribution from secondary formation in the YRD is smaller than that in the US, accounting for about 70% in summer and 40%–54% in winter. The good correlation between total HCHO and secondary HCHO results in a good correlation between total HCHO and RO₂, which means that total HCHO is an appropriate proxy for VOC reactivity in this case. Total HCHO can be regarded as the proxy for VOC reactivity over a 3-year study period from 2015 to 2017 due to the good correlation between total HCHO and secondary HCHO in Nanjing, Hangzhou, and Shanghai ($R \geq 0.59$). If the study period was restricted to 2015, using the ambient HCHO concentration to indicate VOC reactivity would cause significant errors, as ambient HCHO showed a stronger correlation with primary emission than secondary formation during this period (e.g., in the industrial zone of Nanjing and in the suburbs of Shanghai). In addition, the correlation between total HCHO and secondary HCHO depends on season. In winter, the ambient HCHO concentration cannot be represented by the concentration from secondary formation due to the poor correlation ($R < 0.12$, $P > 0.05$). However, in spring (at RJR and DSL), in summer (at RJR, CXT, HK, and PDNA), and in autumn (at RJR and MGQ) total HCHO can be used as a proxy for VOC reactivity ($R > 0.59$, $P < 0.01$).

4.2 HCHO control measures adapted to local conditions

Sources of ambient HCHO at different sites varied largely and were closely related to the types of industries present (Zheng et al., 2016). In order to better adapt HCHO control measures to local conditions, it is meaningful to esti-
Figure 9. The time series of absolute (a, c, e) and relative (b, d, f) contributions of primary sources, secondary sources, and background sources to the HCHO concentration from December 2014 to November 2017 at the HK (a, b), PDNA (c, d), and DSL (e, f) sites in Shanghai. Black dashed lines represent the threshold value for HCHO pollution.

mate the different sources of ambient HCHO. The results in Sect. 3.3 indicate that the main influencing factor for the ambient HCHO concentration depends on the season. Here we define HCHO pollution using a HCHO concentration threshold value of 10 ppbv, which is the 95th percentile of all of the observation data. HCHO pollution events mostly occurred in summer, accounting for 81.0%, 81.3%, and 95.2% of all observed pollution events in Nanjing, Hangzhou, and Shanghai, respectively. In summer, HCHO pollution mostly occurred during the periods from 24 to 27 July 2016 and 22 to 27 July 2017 in Nanjing and from 23 to 27 July 2016 and from 20 to 27 July 2017 in Hangzhou. HCHO pollution in Shanghai lasted longer than episodes in Nanjing and Hangzhou: from 10 to 27 July of 2016 and 2017. Moreover, HCHO pollution events were also observed during the period from 23 to 27 August 2015 and from 5 to 24 August 2017 north of the center of Shanghai. In the Shanghai suburbs, three HCHO pollution periods were also observed: (1) from 19 June to 5 July 2015, (2) from 19 to 23 August 2016, and (3) from 4 to 7 August 2017. During HCHO pollution events, secondary formation contributed most to the ambient HCHO concentration and increased more significantly than the primary HCHO contribution. Compared with the concentrations before pollution days, secondary HCHO and primary HCHO increased by 0.90 and −0.12 ppbv on average in Nanjing, respectively, by 1.33 and 0.21 ppbv in Hangzhou, respectively, and by 1.26 and 0.14 ppbv in Shanghai, respectively. Therefore, decreasing VOC emissions, the source of secondary HCHO, would have a more significant effect on controlling HCHO pollution in Nanjing, Hangzhou, and Shanghai.

The industries of Nanjing include the chemical industry, steel plants, and the cement industry (Zheng et al., 2016). In the Nanjing industrial zone, secondary sources of HCHO were about 4 times larger than primary sources in 2017. Therefore, at industrial sites, decreasing VOC emissions from chemical-related activities is an efficient way to control the ambient HCHO concentration. In Hangzhou, labor-intensive industries are the most important parts of industry economics (Zheng et al., 2016). Secondary HCHO concentrations in the eastern suburbs of Hangzhou were highest
from 2016 to 2017 (Table 1), corresponding to a previous study by B. Lu et al. (2018) which indicated that the largest amount of VOC emissions were measured in the eastern suburbs of Hangzhou and that VOC emissions from furniture manufacturing were the largest source. Secondary HCHO contributed most to ambient HCHO during the period when it was not effectively controlled (from 2015 to 2017). Considering that industrial manufacturing is developing rather rapidly and that VOC emissions occur from many types of industries, i.e., textile industry, printing, leather manufacturing, and shoemaking (Liu et al., 2008; Khan and Malik, 2014), controlling VOC emissions in Hangzhou will be a huge challenge. Unlike Nanjing and Hangzhou, tertiary industry has played a leading role in Shanghai’s economy since 2013, and primary and secondary industries have decreased since 1999 (Chen et al., 2016). However, the HCHO concentration from secondary sources has not been controlled effectively. Decreasing private cars on the road, developing evaporative control regulations, and improving traffic management, which needs to be better suited to local conditions, would help reduce not only VOC emissions but also HCHO emissions, especially in the downtown area (Wang and Zhao, 2008; Liu et al., 2015, 2017).

5 Summary

Tropospheric HCHO VCDs derived from OMPS observations were validated using ground-based FTS measurements in Hefei, from 2015 to 2017. The HCHO VCDs observed by OMPS and FTS were in good agreement ($R = 0.78$). Serious HCHO pollution was observed in the YRD region, especially in southwestern Jiangsu Province (e.g., Nanjing, Changzhou, Wuxi, and Suzhou cities), in northern Zhejiang Province (e.g., Hangzhou, Jiaxing, Huzhou, and Ningbo cities), and in Shanghai. Estimating different sources of ambient HCHO and then formulating control measures adapted to local conditions are effective ways of controlling HCHO pollution. However, measurements of HCHO are scarce, both spatially and temporally. In this study, we analyzed primary and secondary contributions to ambient HCHO using a multiple linear regression based on HCHO observed by OMPS. At the MGQ site in Nanjing, the average relative contributions from primary emissions, secondary formation, and background sources to HCHO from 15 April to 1 May 2015 were 50.5%, 30.8%, and 18.7%, respectively, which was in good agreement with previous research. Therefore, tropospheric HCHO columns observed by OMPS can be used to estimate different sources of ambient HCHO.

Primary and secondary contributions to ambient HCHO were determined in Nanjing, Hangzhou, and Shanghai from 2015 to 2017. Overall, the HCHO concentration from secondary formation contributed most to ambient HCHO in the megacities in the YRD region and influenced the variation of the ambient HCHO concentration in Hangzhou, Shanghai, and urban regions of Nanjing. In comparison, at industrial sites of Nanjing, the annual average HCHO concentration from primary emissions decreased significantly from 2015 to 2016 due to energy saving methods and the reduction of energy consumption, and the contribution from secondary formation to the variation of the HCHO concentration was less significant. At suburban regions in Hangzhou, primary emissions showed a small increasing trend from 2015 to 2017, whereas at urban sites, primary emissions of HCHO in Shanghai were larger than those in Hangzhou. Seasonally, secondary HCHO reached a maximum in summer and a minimum in winter. In spring, summer, and autumn, secondary formation played a crucial role in the variation of the HCHO concentration in urban areas of Nanjing, Hangzhou, and Shanghai, whereas in winter, primary emission contributed more to variation of ambient HCHO in Nanjing, Shanghai, and urban areas of Hangzhou. These findings contribute to formulating effective HCHO pollution control measures. Furthermore, the usability of total HCHO as a proxy of VOC reactivity depends on the timescale utilized. Total HCHO can be regarded as a proxy for VOC reactivity over a 3-year study period, whereas only secondary HCHO can be used as a proxy for VOC reactivity over a shorter study period, e.g., in winter. Therefore separating the HCHO contributions from different sources is of great significance in sensitivity studies of O$_3$ production.

Data availability. The data used in this paper is available on request from the corresponding author (chliu81@ustc.edu.cn and qhhu@iofm.ac.cn).

Supplement. The supplement related to this article is available online at: https://doi.org/10.5194/acp-19-6717-2019-supplement.

Author contributions. WS and CL contributed equally to this study. CL and QH designed and supervised the study, and CL also helped retrieve the satellite data. WS wrote the paper. SZ, YS, and WW retrieved the FTS data, and YZ ran the WRF-Chem model. JL supported the project, and JK contributed to discussions regarding the results.

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

Acknowledgements. This work was supported by National Key Research and Development Program of China (grant nos. 2017YFC0210002, 2018YFC0213104, 2018YFC0213100, and 2016YFC0203302), the National Natural Science Foundation of China (grant nos. 41722501, 91544212, 51778596, and 41575021), and the National High-Resolution Earth Observation Project of China (grant no. 05-Y20A16-9001-15/17-2). The authors
acknowledge the OMPS project for making the OMPS-NPP Nadir Mapper Earth View Level 1B data product available online. We also thank the Ministry of Environment Protection of the People's Republic of China for making data measured by the atmospheric environment automatic monitoring stations available online.

Review statement. This paper was edited by Min Shao and reviewed by two anonymous referees.

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