Significant increase of surface ozone at a rural site, north of eastern China

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Received: 6 September 2015 – Published in Atmos. Chem. Phys. Discuss.: 13 November 2015
Revised: 11 March 2016 – Accepted: 14 March 2016 – Published: 24 March 2016

Abstract. Ozone pollution in eastern China has become one of the top environmental issues. Quantifying the temporal trend of surface ozone helps to assess the impacts of the anthropogenic precursor reductions and the likely effects of emission control strategies implemented. In this paper, ozone data collected at the Shangdianzi (SDZ) regional atmospheric background station from 2003 to 2015 are presented and analyzed to obtain the variation in the trend of surface ozone in the most polluted region of China, north of eastern China or the North China Plain. A modified Kolmogorov–Zurbenko (KZ) filter method was performed on the maximum daily average 8 h (MDA8) concentrations of ozone to separate the contributions of different factors from the variation of surface ozone and remove the influence of meteorological fluctuations on surface ozone. Results reveal that the short-term, seasonal and long-term components of ozone account for 36.4, 57.6 and 2.2 % of the total variance, respectively. The long-term trend indicates that the MDA8 has undergone a significant increase in the period of 2003–2015, with an average rate of $1.13 \pm 0.01$ ppb year$^{-1}$ ($R^2 = 0.92$). It is found that meteorological factors did not significantly influence the long-term variation of ozone and the increase may be completely attributed to changes in emissions. Furthermore, there is no significant correlation between the long-term O$_3$ and NO$_2$ trends. This study suggests that emission changes in VOCs might have played a more important role in the observed increase of surface ozone at SDZ.

1 Introduction

Tropospheric ozone (O$_3$) plays a key role in the oxidizing capacity of the atmosphere (Penkett, 1988) and acts as a greenhouse gas in terms of radiative forcing at the Earth’s surface (IPCC, 2013). Moreover, it is an important precursor of the OH radical, hence changes in its abundance can exert indirect radiative forcing by altering the lifetimes of some other greenhouse gases. Tropospheric O$_3$ originates from photochemical production within the troposphere and the downward transport of stratospheric O$_3$ (Cooper et al., 2014; Monks et al., 2015). Ground-level O$_3$ is subject to in situ chemical reactions and physical processes and is directly affected by precursor emissions, temperature, solar radiation and other meteorological factors.

Observations (Oltmans et al., 2006) and model simulations (Hauglustaine and Brasseur, 2003) indicate that ground-level O$_3$ increased distinctly at northern mid-latitudes during the latter half of the 20th century, which is qualitatively in agreement with the increasing anthropogenic emissions of precursors. Anthropogenic emissions of O$_3$ precursors have been declining in Europe and North America while increasing in East Asia (Streets et al., 2001; Granier et al., 2011). The largest increase in NO$_x$ (NO + NO$_2$) emissions is found in China and appears to have continued into the 21st century based on emission inventories (Streets et al., 2001; Richter et al., 2005; Ohara et al., 2007; Mijling et al., 2013; Kurokawa et al., 2013). From 1990 to 2010, surface O$_3$ in different regions showed inconsistent trends. In the eastern US, sur-
face \(O_3\) was found to decrease strongly in summer, remain largely unchanged in spring and increase in winter, while \(O_3\) increases in the western US were the strongest in spring (Monks et al., 2015). In East Asia, surface \(O_3\) was generally found to be increasing (Cooper et al., 2014). For example, ground-level \(O_3\) in the Northeast Asia region, e.g., Japan (Lee et al., 1998) and Hong Kong (Chan et al., 2003; Wang et al., 2009) increased significantly from the 1990s to 2000s. Enhanced variability of surface \(O_3\), particularly an increasing trend in the monthly highest 5% of the ozone mixing ratios, was reported for the Yangtze River Delta region in eastern China (Xu et al., 2008).

Dynamic factors may contribute to the long-term variations of surface \(O_3\). For example, the long-term increase of surface \(O_3\) was found to be related to the variability in the stratosphere-to-troposphere transport of \(O_3\) (Ordonez et al., 2007; Hess and Zbinden, 2013; Lin et al., 2015a) and changes in transport patterns (Pausata et al., 2012). Decadal circulation shifts played a key role in the autumnal ozone increase and the absence of spring ozone change measured at the Mauna Loa Observatory (3.4 km altitude) over the subtropical Pacific Ocean in the period of 1974–2012 (Lin et al., 2014). However, some studies (Brown-Steiner and Hess, 2014) suggested that the long-term increase in \(O_3\) in some areas mainly to the transport of \(O_3\) and related pollutants from the continental China, where the emissions of \(O_3\) precursors (\(NO_x\) and \(VOC\)) have steadily increased (Ohara et al., 2007; Kurokawa et al., 2013). Studies by Lin et al. (2015b) indicate that mean springtime ozone level over western North America in the most recent decade has increased by 5.9 ± 2.1 ppbv compared to the 1980s, which could be attributed to part of the rising Asian ozone precursor emissions and global methane. Because of increasing emissions of \(O_3\) precursors due to the sustained economic growth in China (Zhang et al., 2007), fueled by favorable photochemical conditions, China has likely been experiencing severe photochemical pollution. Due to this industrialization, increases are expected in some regions. Limited studies have indicated this is indeed the case. Ding et al. (2008) analyzed \(O_3\) data from the MOZAIc (measurement of ozone and water vapor by Airbus in-service aircraft) program and obtained a 2% year\(^{-1}\) rate of increase in the daytime \(O_3\) in the lower troposphere over Beijing and its surrounding areas for the period of 1995–2005. Xu and Lin (2011) analyzed the TOR (tropospheric ozone residue) data over the period 1979–2005 and found a significant upward trend in tropospheric \(O_3\) over the North China Plain for all seasons except for winter, with a maximum rate of increase of 1.10 DU per decade for summer. Wang et al. (2009) found that surface \(O_3\) at a regional station in Hong Kong increased at an average rate of 0.58 ppbv year\(^{-1}\) from 1994 to 2007 and they associated the trend with an increase in tropospheric \(NO_2\).

Information regarding the trends in surface \(O_3\) concentrations in the different regions of China, particularly those with high anthropogenic emissions rates, is urgently needed. Due to a lack of long-term observations, it is difficult to gain reliable results about the long-term trends of surface \(O_3\) in various regions of China. Recently, the long-term trend of surface \(O_3\) in western China was reported by Xu et al. (2015), based on the observations at the Mt. Waliguan baseline station. So far, there has been no report of changes of surface \(O_3\) levels in highly polluted eastern China during the recent decade. In this paper, we present trends in surface \(O_3\) in the North China Plain, based on the measurements from a rural site. The relative contributions of meteorological factors and anthropogenic emissions are investigated, which provide further insight into the potential causes of the observed trend of surface \(O_3\).

2 Data and methods

2.1 Site and measurements

Surface \(O_3\) and ancillary data were collected at the Shangdianzhi (SDZ, 40.65° N, 117.10° E, 293.3 m a.s.l.) station. SDZ is one of the regional Global Atmosphere Watch (GAW) stations, located about 100 km northeast of suburban of Beijing. The 30 km radius surrounding the site contains only small villages with a sparse population and insignificant anthropogenic emission sources. The observation facilities are situated on the south slope of a hill, which is surrounded by mountainous areas except in the southwest sector. Fruit trees and corn are grown in the fields surrounding the site. Previous studies (Lin et al., 2008; Xu et al., 2009) suggest that the observations of pollutants at SDZ reflect the regional-scale air quality of North China.

The maximum daily average 8 h (MDA8) concentrations of \(O_3\) were calculated from hourly averages of \(O_3\) from October 2003 to June 2015 and are used in the following analysis. To facilitate the analysis, ambient \(NO_2\) concentration and temperature measured at SDZ in the surface layer during the same time period were processed to obtain daily averages. Details of the observations and the quality assurance and quality control (QA/QC) procedures were described by Lin et al. (2008).

2.2 Analysis methods

It is well known that meteorology plays an important role in ozone formation and transport. Ground-level ozone concentrations are strongly influenced by fluctuations of meteorological parameters. Therefore, it is difficult to distinguish the trend of ozone related to the change in emissions from that related to meteorological impacts. In order to filter out or minimize the influence of meteorology on ozone levels, a method called Kolmogorov–Zurbenko (KZ) filter (Rao and Zurbenko, 1994) is used to separate data into short-term, seasonal and long-term variations. The KZ filter is based on an

iterative moving average that removes high frequency variations in the data. The method is briefly described below.

The KZ(m, n) filter is defined as n applications of a moving average of m points. The moving average can be expressed as

\[ Y_i = \frac{1}{m} \sum_{j=-k}^{k} X_{i+j}, \]  

where \( m = 2k + 1 \), and the calculated \( Y_i \) becomes the input for the second pass, and so on.

Data filtered by the KZ filter preserve information related to physical processes, whereas data treated by some other techniques may remove unwanted information, but at the same time distort phenomena of interest. Eskridge et al. (1997) compared the KZ filter method with several others, such as wavelet transform, anomalies, etc. and demonstrated that the KZ filter has the same level of accuracy as the wavelet transform method. In addition, the magnitude of the long-term trend estimated by the KZ filter provides estimates with approx. 10 times higher confidence than the other methods. However, the width moving average (m) of the KZ filter with wide windows will dampen sharp breaks of variations. Based on the KZ filter, an adaptive filter was developed by Zurbenko et al. (1996), which dynamically adjusts the width of m according to the rate of change of the process. As the rate of change increases, the m decreases. The modified KZ filter method is applied in this paper. More details on this method can be found in Zurbenko et al. (1996).

Rao et al. (1997) developed a method to separate different phenomena present in time series of both meteorological and ozone data, which have different characteristics such as long-term and short-term variations. Following the method, it is assumed that the time series of ozone can be partitioned as

\[ O(t) = W(t) + S(t) + e(t), \]  

where \( O(t) \) is the original time series, \( W(t) \) is the mesoscale and synoptic-scale variation, \( S(t) \) is the seasonal change, \( e(t) \) is the long-term (trend) component. Rao et al. (1997) found that when KZ15,5 and KZ365,3 filters are applied to the raw data, several influences could be removed and the actual variation of ozone at different scales would be obtained. \( W(t) \), \( S(t) \) and \( e(t) \) can be calculated using the following formulas.

\[ W(t) = O(t) - KZ_{15,5}, \]  
\[ S(t) = KZ_{15,5} - KZ_{365,3}, \]  
\[ e(t) = KZ_{365,3} \]

The Rao et al. (1997) method was implemented in this work.

Figure 1. Percentile-box plot of yearly statistics of MDA8 values of ozone at SDZ, in the period of 2004–2015. Boxes depict interquartile range and median; the squares depict the mean; whiskers depict 10th and 90th percentile; dots depict maximum values.

3 Results and discussion

3.1 General characteristics of yearly ozone distribution

Yearly MDA8 statistics were calculated for 2004–2015 and are presented in Fig. 1. Since the ozone observation at SDZ commenced in October 2003, no reliable yearly MDA8 statistics can be obtained. It is noted that data from 2015 cover only the first 6 months. Although only the first 6 months records in 2015 are used for the statistics, the maximum of the MDA8 values in this year exceeded 160 ppb, only second to that in 2012. The yearly average of MDA8 varied from 49.3 to 60.2 ppb during 2004–2014, with a highly significant positive trend (1.05 ± 0.14 ppb year\(^{-1}\), \( R = 0.93, P < 0.0001 \)). We also observed a similar fluctuation of the median value within the range of 43.3 to 53.0 ppb, with a positive trend (0.62 ± 0.20 ppb year\(^{-1}\), \( R = 0.72, P < 0.05 \)) from 2004 to 2014. The MDA8 level was relatively stable during 2004–2006, with a maximum of approx. 120 ppb. However, the annual maximum value exhibited a dramatic increase from 123 ppb in 2006 to 165 ppb in 2015. This increase coincided with an increase of in the size of the vehicle fleet in eastern China. For example, in Beijing, the number of registered vehicles was 2.30 million in 2004, 2.88 million in 2006, 4.81 million in 2010, and 5.60 million in 2014 (http://www.bjjgl.gov.cn/jgj/ywsj/index.html). Both the maximum O\(_3\) value and the vehicle number increased dramatically in the period 2004–2015. Nevertheless, it is not possible to derive a reliable long-term trend in the median or maximum value solely from the data shown in Fig. 1, nor can we directly attribute the observed changes in surface O\(_3\) to the increase in vehicles.
3.2 Ozone time series separated by KZ filter

The ozone time series (MDA8 value) from the SDZ site was separated using the method described in Sect. 2.2. Figure 2 shows the original time series of MDA8 values (Fig. 2a) and the time series of the separated short-term, seasonal and long-term components (Fig. 2b–d). The original MDA8 exhibits a distinct seasonal variation, with an overlapping of high-frequency noises (Fig. 2a). Removing the short-term component (Fig. 2b) leads to clearer seasonal cycles shown in Fig. 2c. As can be seen in Fig. 2c, there are evident double peaks of ozone during summer in each year, which are not so obvious in the original time series (Fig. 2a). Generally, the double peaks occur in June and September, respectively, and the dip in between occurs in July or August when relatively abundant rainfall damps ozone formation and accumulation. Under the influence of the summer Asian monsoon, rainfall in July and August at SDZ can amount to more than 40% of the annual rainfall. Figure 2c also demonstrates some irregularities in the seasonal cycle, particularly the year-to-year changes in the levels of annual maximum, annual minimum and the dip between the double peaks. The seasonal fluctuations have to be accurately removed to get the long-term trend, as data for the trend analysis are required to be independent of season and normally distributed. The short-term component (Fig. 2b) showed high frequency variations, ranging between −60 and 70 ppb, which are composed of noise (or fluctuation) caused by mesoscale and synoptic-scale meteorological processes. Synoptic-scale events have a timescale from 2 days to 3 weeks, which could be removed by smoothing with the KZ filter for a window size of 15 days and five iterations. To further illustrate the short-term component, a quantile–quantile ($Q–Q$) plot of $W(t)$ is presented in Fig. 3. The $Q–Q$ plot indicates that $W(t)$ basically obeys a normal distribution, with a mean value of 0.002 ppb, suggesting that the KZ$_{15,5}$ filter effectively removed $W(t)$ from $O(t)$.

Through the previous steps and using the Eqs. (2)–(5), we obtained the long-term trend of MDA8 at SDZ, as shown in Fig. 2d. This long-term trend reveals a rapid increase of the daily high value of surface ozone at the SDZ site in the last decade. It is noteworthy that the increase is not at a stable rate but with large inter-annual variations. Linear regression (not shown) indicates that the average increase rate during 2004–2015 was $1.13 \pm 0.01$ ppb year$^{-1}$ ($r^2 = 0.92$). Previous work by Ding et al. (2008) using MOZAIC data obtained a yearly increase of 2% (about 1 ppb year$^{-1}$) of O$_3$ in the boundary layer around Beijing in the period of 1995–2005, which agrees well with our result. Therefore, the greater Beijing area, probably the North China Plain, has been suffering a rapid ozone increase for the last 2 decades.

In view of the air pollution problems, the central government of China issued a revised National Ambient Air Quality Standard (CNAAGS, GB 3095-2012) in 2012, which has taken effect across the country since 1 January 2016 and sets the MDA8 O$_3$ limits to 100 µg m$^{-3}$ (46.7 ppb) and 160 µg m$^{-3}$ (74.7 ppb) for national reserve areas and residence/commercial areas, respectively. As can be seen in Fig. 2a, O$_3$ exceedance would occur quite often in the warm seasons if the new CNAAGS had been implemented.

We also examined the contributions of different components to the total variance of MDA8, which was calculated from the unfiltered data. The contributions of the short-term and seasonal components to the total variance are about 36.4 and 57.6%, respectively. The long-term component accounts for only 2.2% of the total variance. The covariance terms amount to less than 4% of the total variance, indicating an effective separation of different components. The long-term
component makes only a much smaller contribution than the other two components, confirming the necessity to clearly separate the short-term and seasonal variations from the data to obtain the long-term trend.

3.3 Cause analysis

The long-term trend of the ozone concentration can be caused by the changes of both pollutant emissions and related meteorological variables. Climate variability and circulation shifts may lead to long-term changes of $O_3$ as discussed in Lin et al. (2014; 2015a, b). To assess the influence of precursor emissions on the ozone trend, the meteorological and chemical impacts have to be separated. However, both meteorological and chemical impacts are complicated, not to mention the interactions among meteorology, precursor emissions and photochemical reactions. Therefore, a clear separation of meteorological and chemical impacts is hardly possible purely based on observational data. Nevertheless, apportionment of the $O_3$ trend to precursor emissions and other causes are worthy of further study.

Although many meteorological variables can influence the photochemical formation of $O_3$, the prevailing variable is temperature. The increase of temperature can speed up the photochemical reactions, strengthen the emissions of biogenic VOCs and reduce wind speed, among others (Lin et al., 2001; NRC, 1991; Pusede et al., 2015). In certain regions, temperature is also closely related to the intensity of solar radiation, which plays a critical role in the photochemical formation of $O_3$. Thus, we took temperature as a key meteorological parameter and investigated the relationship between $O_3$ and temperature, with the hope to obtain the influence of emission changes on the long-term trend of $O_3$. The initial step was to divide the time series of temperature into three components in Eq. (2), just as done for that of MDA8 (Fig. 2). The results of the different components of temperature are given in Fig. 4. Unlike the trend of MDA8 of $O_3$, the long-term component for temperature at SDZ shows a slight decrease trend ($R^2 = 0.015$) (Fig. 4d) and this long-term component accounts only for 0.16 % of the total variance of temperature.

The unfiltered data of $O_3$ and temperature are less correlated ($R^2 = 0.50$, $P < 0.0001$), presumably due to the strong influence of the short-term component. Figure 5 compares the derived seasonal cycles of the daily mean temperature (from Fig. 4c) and the MDA8 of $O_3$ (from Fig. 2c). A similar-
ity is evident between both seasonal cycles. However, there is also a distinct phase lag of the seasonal cycle between \( O_3 \) and temperature, due to the influence of other processes on the \( O_3 \) level. Rao et al. (1995) found a similar phase lag of about 3 weeks in the data from the northeastern United States. In our case, the linear correlation between \( O_3 \) and temperature becomes the strongest \( (R^2 = 0.83, P < 0.0001) \) when the temperature data are lagged by 17 days (Fig. 6).

When only considering the influence of temperature, the seasonal and long-term components of \( O_3 \) could account for 93 % of the total variance at the Cliffside Park, New Jersey (Rao and Zurbenko, 1994), while in our case, it accounts for just 83 % (see \( R^2 \) in Fig. 6). We tried to add more meteorological factors that could affect \( O_3 \) production, such as solar radiation and relative humidity. However, the correlation was improved by no more than 0.5 %. This implies that the changes in emissions might have a more important influence on surface \( O_3 \) at SDZ than at Cliffside Park. This view is consistent with the rapid increases in anthropogenic emissions in China (particularly the North China Plain) during the last decade (Mijling et al., 2013).

Assuming that the residual of the total variance of \( O_3 \), after subtracting the contribution related with temperature, was all caused by pollutant emissions, the long-term trend of \( O_3 \) attributable to changes in emissions, can be determined by performing a linear regression between time and the noise-free, temperature-independent \( O_3 \) values \( (\varepsilon(t)) \), which are derived using Eq. (6):

\[
O_{3,\text{fit}}(t) = aT_{3,\text{fit}}(t + 17) + b + \varepsilon(t),
\]

where \( O_{3,\text{fit}}(t) \) is the filtered \( O_3 \) concentration, \( T_{3,\text{fit}}(t + 17) \) is the filtered temperature lagged by 17 days, \( a \) and \( b \) are fitted parameters, \( \varepsilon(t) \) is the residual of the relationship. Here, \( \varepsilon(t) \) reveals the changes in ozone attributable to changes in emissions.

Figure 7 shows the time series of the noise-free and temperature-independent \( O_3 \), which is basically equal to the long-term component of \( O_3 \) only under the influence of emission changes. Most of the data in Fig. 7 are within a 95 % confidence interval band except for some special cases in the summer months. In summer, temperature is not the dominant restricting factor for \( O_3 \) production compared to other factors, such as rainfall and precursor concentrations. Substantial negative influences occurring in 2005 and 2006 can be explained by stronger impact of Asian summer monsoon on surface \( O_3 \) (Lin et al., 2008). The results in Fig. 7 indicate that the influence of emission has been varying substantially but with an average increase rate of \( 1.19 \pm 0.03 \) ppb year\(^{-1}\). This increase rate is very close to the average long-term trend of MDA8 of \( O_3 \) \((1.13 \pm 0.01 \) ppb year\(^{-1}\)) in Fig. 2d, implying that the increase of \( O_3 \) in the period of 2003–2015 could be mainly attributed to the emission changes and that the meteorological factors had only a tiny negative influence. Jaffe and Ray (2007) also found that the temperature change had little influence on long-term ozone trends in the western US.

Some studies suggested that the trends of surface \( O_3 \) at the similar latitude as SDZ could be attributed partly to the reduced titration by NO (Chou et al., 2006; Itano et al., 2007). In order to assess the effect of changing NO titration on the long-term trend of \( O_3 \), we examined the long-term measurements of \( NO_2 \) at SDZ in the period of 2004–2015. A comparison of the long-term trend of \( O_3 \) with that of \( NO_2 \), which was also extracted using the previous methods, is displayed in Fig. 8. The evolution of the \( NO_2 \) trend can be divided into three stages, i.e., a substantial decrease of \( NO_2 \) occurring during the first 3 years, followed by a small increase in the period of 2007–2010 and finally a gradual decrease in the period of 2011–2015. The large decrease of \( NO_2 \) in the period of 2004–2006 corresponded to the control of coal consumption around Beijing, especially for the Olympic Games in 2008 (Zhang et al., 2010; Gao et al., 2011) and to the relocation of the Capital Steel and Iron Company, which was one of the largest steel producers in China.
of the largest industrial sources in Beijing. The NO\textsubscript{2} increase from 10.2 to 13.5 ppb between 2007 and 2010 corresponded with the rapid increase in the number of vehicles in Beijing from 3.1 million to 4.8 million (http://www.bjctbj.gov.cn/igj/ywysj/index.html). From 2011 to 2015, the new standards for vehicle emissions and measures for reduction of NO\textsubscript{2} emission from power plants were implemented, which may have helped to reduce the NO\textsubscript{2} concentration. The long-term trends of O\textsubscript{3} and NO\textsubscript{2} given in Fig. 7 do not show any coincidence. Therefore, it is nearly impossible that the reduced NO\textsubscript{2} titration had led to the increase of surface O\textsubscript{3} at SDZ. Previous studies (Ge et al., 2010; 2012) showed that the ozone production efficiency at SDZ varied in values from 0.2 to 21.1, with an average of 4.9, implying that ozone formation at SDZ could be more sensitive to VOCs than to NO\textsubscript{2}. Accordingly, we believe that the changes of VOC emissions and the ratio of VOCs to NO\textsubscript{2} might have caused the increase of surface O\textsubscript{3} observed at SDZ. Unfortunately, no systematic VOC observations are available from the SDZ site so that we cannot prove this supposition conclusive. However, a large increase in the anthropogenic emissions of non-methane hydrocarbon (NMHC) can be inferred from the Multiresolution Emission Inventory for China (MEIC) (http://www.meicmodel.org) for Beijing in the period of 2004–2012, which supports our view, although the emission data are questioned by a recent study (Wang et al., 2015).

4 Summary

We separated the time series of maximum daily average 8 h (MDA8) concentration of surface O\textsubscript{3} observed at SDZ in the period of 2003–2015 into various spectral components using a modified KZ filter. This separation has led to a better understanding of the variation of surface O\textsubscript{3} at the site and its relationships with the meteorological and precursor variables, enabling us to unravel the trend of O\textsubscript{3} from the original data containing noises and seasonality, and to estimate the contribution of changes of precursor emissions to the trend. Our analysis reveals that the short-term, seasonal and long-term components of O\textsubscript{3} data from the SDZ site accounted for 36.4, 57.6 and 2.2 % of the total variance, respectively.

It is found that the MDA8 of O\textsubscript{3} at the site north of eastern China has undergone a significant increase in the period of 2003–2015, at an average rate of 1.13 ± 0.01 ppb year\textsuperscript{−1}. Together with the reported yearly increase rate of 2 % in the lower tropospheric O\textsubscript{3} around Beijing in the period of 1995–2005 (Ding et al., 2008), we conclude that the north part of eastern China (i.e., the North China Plain) may have been suffering a rapid increase in the O\textsubscript{3} level for at least 2 decades. By eliminating the influence of air temperature, we found that the observed increase of surface O\textsubscript{3} in the period of 2003–2015 was mainly induced by the emission changes, and the meteorological factors exerted only a tiny negative influence. Our result also indicates that the change of VOC emissions might have played a more important role in the O\textsubscript{3} increase than the effect of NO\textsubscript{2}.

Because fine particulate pollution has been very severe in eastern China, the central government of China has implemented several measures to control PM\textsubscript{2.5} pollution, including reductions of both NO\textsubscript{2} and VOCs. This, however, risks further O\textsubscript{3} increases as VOCs to NO\textsubscript{2} ratio more favorable to ozone production may be reached. Thus, further studies are needed to trace the ozone trend and its influence in eastern China.

Acknowledgements. The authors would like to thank the staff of the Shangdianzhi station for their excellent work. This research is supported by the National Science Foundation of China (41475135, 41330422), Beijing Natural Science Foundation (8132025, 8152018), CCSSF201505. We also thank the referees and editor for their constructive suggestions.

Edited by: J. B. Burkholder

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