Impact of pollution controls in Beijing on atmospheric oxygenated volatile organic compounds (OVOCs) during the 2008 Olympic Games: observation and modeling implications

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Abstract. Oxygenated volatile organic compounds (OVOCs) are important products of the photo-oxidation of hydrocarbons. They influence the oxidizing capacity and the ozone-forming potential of the atmosphere. In the summer of 2008, 2 months of emission restrictions were enforced in Beijing to improve air quality during the Olympic Games. Observational evidence reported in related studies that these control measures were efficient in reducing the concentrations of primary anthropogenic pollutants (CO, NOₓ and non-methane hydrocarbons, i.e., NMHCs) by 30–40%. In this study, the influence of the emission restrictions on ambient levels of OVOCs was explored using a neural network analysis with consideration of meteorological conditions. Statistically significant reductions in formaldehyde (HCHO), acetaldehyde (CH₃CHO), methyl ethyl ketone (MEK) and methanol were found to be 12.9, 15.8, 17.1 and 19.6%, respectively, when the restrictions were in place. The effect of emission controls on acetone was not detected in neural network simulations, probably due to pollution transport from surrounding areas outside Beijing. Although the ambient levels of most NMHCs were reduced by ~35% during the full control period, the emission ratios of reactive alkenes and aromatics closely related to automobile sources did not present much difference (≤30%). A zero-dimensional box model based on the Master Chemical Mechanism version 3.2 (MCM3.2) was applied to evaluate how OVOC production responds to the reduced precursors during the emissions control period. On average, secondary HCHO was produced from the oxidation of anthropogenic alkenes (54%), isoprene (30%) and aromatics (15%). The importance of biogenic sources for the total HCHO formation was almost on par with that of anthropogenic alkenes during the daytime. Anthropogenic alkenes and alkanes dominated the photochemical production of other OVOCs such as acetaldehyde, acetone and MEK. The relative changes of modeled HCHO, CH₃CHO, methyl vinyl ketone and methacrolein (MVK + MACR) before and during the pollution controlled period were comparable to the estimated reductions in the neural network, reflecting that current mechanisms can largely explain secondary production of those species under urban conditions. However, it is worth noting that the box model overestimated the measured concentrations of aldehydes by a factor of 1.4–1.7 without consideration of loss of aldehydes on aerosols, and simulated MEK was in good agreement with the mea-
measurements when primary sources were taken into consideration. These results suggest that the understanding of the OVOCs budget in the box model remains incomplete, and that there is still considerable uncertainty in particular missing sinks (unknown chemical and physical processes) for aldehydes and absence of direct emissions for ketones.

1 Introduction

Oxygenated volatile organic compounds (OVOCs), such as aldehydes, ketones and alcohols, are mostly produced in the atmosphere by the oxidation of biogenic, anthropogenic hydrocarbons and other organic species (Finlayson-Pitts and Pitts, 2000), and also directly emitted by vegetation (Park et al., 2013), biomass burning (Yokelson et al., 2009; Mason et al., 2001; Andreae and Merlet, 2001), fossil fuel combustion (Schauer et al., 1999, 2002) and industries (Singh et al., 1994). They are lost through oxidation by OH, photolysis, and deposition/surface uptake. Aircraft measurements showed that the vertical distribution (0–12 km) of the total concentrations of oxygenated organics was 2–5 times as abundant as the sum of C2–C8 non-methane hydrocarbons (NMHCs) at all altitudes in the remote Pacific troposphere (Singh et al., 2001). Formaldehyde (HCHO) and other oxygenated hydrocarbons dominated the total OH loss with VOCs (volatile organic compounds) in clean air masses (Goldan et al., 2004). In polluted atmospheres the interaction of primary and secondary VOCs is closely coupled with the formation of ozone and secondary organic aerosols (SOA). And usually OVOCs serve as intermediate products in these oxidation processes, which helps to estimate the formation potential of ozone (Shao et al., 2011) and other products. And OVOCs also affect the oxidizing capacity of the atmosphere. Some previous studies found that carbonyls are one of the most important radical sources, especially for the wintertime in polluted urban environments (Emmerson et al., 2005). Accordingly, it is essential to understand the sources, sinks and chemistry of OVOCs quantitatively, while the role of OVOCs in heavy polluted regions remains largely unexplored partly due to the lack of reliable observations of them.

Beijing, the capital of China and one of the most populous megacities in the world, is located in northern China. Accompanied with the rapid population growth and economic expansion, severe air pollution in Beijing has attracted global attention. The city has been known for its increasing ozone concentration (Parrish and Zhu, 2009; Shao et al., 2009) and hazy skies (Chan and Yao, 2008). In preparation for the 2008 Summer Olympics and to clean up the city’s air, Beijing imposed a number of pollution control measures during the games. Thus, coordinated observations on such a restricted occasion create a valuable opportunity for studying how OVOCs respond to the emission reductions, testing and refining the current knowledge of formation pathways of secondary OVOCs under polluted conditions.

After Beijing was selected as the host of the 2008 Summer Olympics, the Beijing government started to implement a series of long-term emission control regulations, such as the closure of heavy polluting industries (chemical factories, cement plants, steel manufacturing, etc.) in the southeast of the city, relocating nearly 200 factories out of the Beijing area; also tightening the vehicle emission and fuel quality standards, and accelerating the retirement of old vehicles. More stringent, short-term measures were put into effect from 1 July to 20 September 2008 such as temporarily halting industrial production, suspending construction and evaporative emissions. Nearly 2 million vehicles were banned from the roads step by step (UNEP, 2009): (1) a total of 350,000 yellow-labeled vehicles with high emissions were not allowed to enter the city starting from 1 July; (2) 50% of privately owned vehicles were restricted to run on alternate days (based on their license plate numbers) in the metropolis and 70% of government-owned vehicles were halted from operating from 21 July to 20 September, which is a so-called “full control period”. As a result, the traffic emissions on VOCs, CO, NOx, and particulate matter (PM) were estimated to be reduced by 50% (Wang et al., 2010b).

The 2 months of emission restrictions had a noticeable impact on major pollutants directly emitted from fuel combustions, which have been validated by on-road measurements (Wang et al., 2009a, b; Wang and Xie, 2009), ground-level monitoring (Wang et al., 2010a, c; Chou et al., 2011; Zhang et al., 2009) and satellite-based observations (Witte et al., 2009; Worden et al., 2012; Lyapustin et al., 2011). Satellite measurements over Beijing showed a 43% reduction of tropospheric column NO2 during the games (Witte et al., 2009), and a 32% reduction in CO for 2008 with respect to 2007 was estimated from MOPITT (Measurement of Pollution in the Troposphere) satellite retrievals (Worden et al., 2012). Wang et al. (2009a) found a dramatic decrease of primary pollutants from on-road measurements, by up to 54% for CO, 41% for NOx and 66% for aromatics (benzene, toluene, ethylbenzene and xylenes, BTEX). And Wang et al. (2010a) reported the averaged mixing ratios of NMHCs near the Olympic Stadium during the games were reduced by 35%, compared with the concentrations measured in June (before the restrictions). The monthly concentrations of NO, NO2 and CO were reduced by 76.8, 29.7 and 27.8% in August 2008 compared to previous years for the same month (Chou et al., 2011). In contrast, the observed mixing ratio of O3 increased by 16% during the full control, compared to the period prior to 20 July (Wang et al., 2010c). The higher levels of ozone were influenced by the declined O3–NO titration and the shift in different O3 formation regimes (Chou et al., 2011), changes in weather conditions, and/or by the transport of photochemical plumes from the surrounding areas to Beijing (Wang et al., 2010c). So, the influence of control measures on secondary pollutants still needs more investigation.
The limited OVOCs data sets available in Beijing suggest that the ambient levels of carbonyls were observed to be 3–5 times higher than the levels in Hong Kong, about 35% of those in Mexico City (Pang and Mu, 2006), and comparable to those in Rio de Janeiro, Brazil (Grosjean et al., 2002). In summer, about 60–70% of aldehydes in Beijing were associated with the photochemical oxidation of anthropogenic and biogenic VOCs, and 10–15% of them were attributed to the primary emission from anthropogenic sources (Liu et al., 2009). For ketones and alcohols, anthropogenic primary emissions were shown to be the largest contributor (Liu et al., 2009; Yuan et al., 2012). The 2008 summer provides the opportunity to estimate the impact of reduced emissions on ambient OVOCs. In this study, firstly a neural network approach was used to link VOC mixing ratios to meteorological conditions, which helps to ascertain that the observed changes in VOCs are significant due to the emission restrictions, rather than variations caused by different meteorological conditions. Secondly, a box model using the Master Chemical Mechanism (MCM) was employed to investigate secondary formation of OVOCs during the full control period, in particular, of two most abundant aldehydes (formaldehyde, acetaldehyde), two ketones (acetone and MEK) and unique oxidation products of isoprene chemistry (MVK + MACR). Moreover, this study also aims to quantify the changes of secondary OVOCs by the existing mechanisms, responding to the reduced precursors (NMHCs) and other pollutants (CO, NO, NO2, O3), and to analyze the similarity or discrepancy between the measured and modeled changes in OVOCs before and during the full control.

2 Methodologies

2.1 Monitoring sites and experimental methods

As part of the CAREBEIJING 2008 campaign, the measurements of VOCs and other air pollutants (CO, NOx, O3, etc) were conducted at an intensive observation site on the campus of Peking University (PKU) from 3 July to 27 August 2008. The PKU site (39.99° N, 116.31° E) was located in the northwest of downtown Beijing, which has been considered to be representative of a typical urban environment with different mixtures of high density traffic, commercial, residential, and electronic companies (Liu et al., 2009; Cheng et al., 2008; Song et al., 2007). The air inlets for the instruments were set up on the top of a six-story building (~25 m a.g.l.). Three online techniques were used to quantify NMHCs and OVOCs during the campaign, including a proton-transfer-reaction mass spectrometer (PTR-MS), a gas chromatograph with mass spectrometer and flame ionization detector (GC-MS/FID) and a Hantzsch fluorimetric monitor.

2.1.1 Online measurements for NMHCs and OVOCs

A high-sensitivity PTR-MS (Ionicon Analytik, Innsbruck, Austria) was used for ambient measurements of selected NMHCs and OVOCs. This technique ionizes VOC molecules by transferring protons (H+) from protonated water ions (H3O+) in a drift tube, followed by a quadropole mass spectrometer detection of the product ions (Lindinger et al., 1998; de Gouw and Warneke, 2007). The air sample was drawn through an 8 m long perfluoroalkoxy (PFA) tube (0.25 in. OD × 0.156 in. ID) into the gas-handling system by a diaphragm pump at the flow rate of 2–3 slpm. A small flow of 220–240 scm was branched from the main flow to the PTR-MS. A 2.0 µm pore-size 47 mm PTFE (polytetrafluoroethylene) filter at the inlet was used to remove particles from the sample flow.

The operating parameters of PTR-MS used herein are similar to those described previously in Yuan et al. (2010a). The instrument was operated at 2.3 mbar drift tube pressure and 600 V drift voltage, 35 masses were sampled with the time resolution of 160 s. Instrument background signals were measured every 35 cycles (~1.5 h) by switching the sample air to an activated charcoal trap for 5 cycles (~13 min). Target ions with the dwell time of 5 s for each cycle included OVOCs such as methanol (m/z = 33), acetaldehyde (m/z = 45), acetone (m/z = 59), MVK + MACR (m/z = 71), methyl ethyl ketone (MEK, m/z = 73); aromatics such as benzene (m/z = 79), toluene (m/z = 93), styrene (m/z = 105), C8 aromatics (m/z = 107) and C9 aromatics (m/z = 121); isoprene (m/z = 69) and acetonitrile (m/z = 42). The PTR-MS was calibrated every week using a commercial TO15 gas standard (Air Environmental Inc. Denver, USA) at seven concentration levels ranging from 1 to 15 ppb. For most species, the detection limits were between 23 and 60 ppptv, except methanol (141 ppptv), acetone (148 ppptv) and MEK (110 ppptv). The relative errors of the PTR-MS measurements decrease with the mixing ratios of target species, as shown in Fig. S1, and the precisions of all detected species were below 5% at the level of 1 ppbv.

An automated GC-MS/FID system with two columns and two detectors was deployed by the Research Center for Environmental Changes (RCEC), Academia Sinica, Taiwan, which was used to measure C2–C4 hydrocarbons and C5–C11 hydrocarbons simultaneously. Technical details for this instrument were presented in Chou et al. (2011). A total of 65 NMHCs were quantified hourly during the full control period (24 July–27 August), the precision for most of the species was within 2% and the limits of detection were below 30 ppbv.

The comparison between the PTR-MS and GC-MS data showed a good agreement for aromatics (benzene, toluene, C8 aromatics and C9 aromatics) during the campaign, as shown in Fig. S2a, with the slopes varying between 0.86 and 1.21 and the correlation coefficients larger than 0.89. The isoprene concentrations measured by PTR-MS were systemati-
cally higher than those measured by GC-MS, particularly for the mixing ratios lower than 1 ppbv, shown in Fig. S2b, possibly because signals at m/z 69 detected by PTR-MS were influenced by some interferences from pentanal, methylbutanal, pentenol (De Gouw et al., 2003) and cycloalkanes (Yuan et al., 2014). It should be noted that the artifact of PTR-MS measurements on isoprene leads to an additional background in anthropogenic-emission-dominated areas. So, the isoprene data from the GC-MS is used in the following model calculations.

HCHO was measured by an online Hantzsch fluorimetric monitor at the time resolution of 10 min. The instrument is described in previous publications (Dasgupta et al., 2005; Li et al., 2010) and was based on sensitive wet chemical fluorimetric detection of HCHO. A diffusion scrubber was designed to strip and collect HCHO from the gas phase into the liquid phase. The detection limit for HCHO was within 100 pptv, and the precision was below 10%.

2.1.2 Measurements for other air pollutants and parameters

A comprehensive set of instruments was installed at the PKU site to monitor the air quality. CO was measured by a non-dispersive infrared sensor (NDIR) with an integration time of 1 min (48 CTLE, Thermo Environmental Instruments, TEI, USA). Zero calibrations were done every 2 h and span calibrations were performed at midnight (00:00–1:00 LT) each day during the campaign. Ozone was measured by a UV absorption detector (Model 49i, TEI). NO was measured using a NO/NO3 chemiluminescence analyzer (Model 42iTL, TEI), with the instrument sequentially measuring NO2 through a photolytic converter.

HONO was measured using an online stripping coil sampler/ion chromatography (IC) system at the time resolution of 15 min. The instrument has a detection limit of 8 ppt with an uncertainty of 7%, other technical details were presented in Cheng et al. (2013).

For aerosol measurements at the site, particle size distributions from 3 to 900 nm (mobility diameter) were measured by a TDMPS (twin differential mobility particle sizer) system with 10 min time resolution, as presented in Wang et al. (2011). An Aerodyne HR-TOF-AMS (high-resolution time-of-flight aerosol mass spectrometer) was deployed to measure chemical compositions of PM1. A detailed description of the system was shown in Huang et al. (2010).

A meteorological station was operated simultaneously to monitor meteorological parameters (temperature, RH, pressure, wind speed and wind direction) on the campus, which was about 400 m away from the air quality intensive site. The photolysis frequencies of O1D and NO2 were measured at the PKU site by two filter radiometers (Bohn et al., 2008) with a time resolution of 5 s. The uncertainties of J values were within 10%.

2.2 Methodology for neural network

The temporal variation of VOC concentrations at a receptor site reflects a combination of changes in emission location and strength, meteorological conditions, chemical loss and secondary formation. Previous studies indicated that meteorological conditions, especially local wind speed and direction, played an important role in shaping the air quality of Beijing (Wu et al., 2008). We cannot quantify the influence of emission restrictions on ambient VOCs by simply comparing their levels during the control period with those before and after. Therefore, following the method introduced by Cermak and Knutti (2009), a feed-forward neural network was used to establish a statistical relationship between VOC mixing ratios and meteorological parameters from the reference data (i.e., before the full control) and then to predict the concentrations given the meteorological conditions during the games, if there were no restriction measures taken. Thus, the difference between the observed and predicted VOC concentrations is helpful to determine the effective changes due to the air quality measures.

The multilayer perceptron (MLP) network in SPSS17.0 was used in this work. As the concentrations of VOCs displayed near lognormal distributions, the natural logarithm of VOC concentration, ln(VOC), was selected as the “dependent variable” of inputs in the MLP network, and the meteorological parameters as “covariates” of inputs, including wind vector (u, v), previous day u, previous day v), precipitation, relative humidity, temperature and pressure. The observed data from 3 July to 20 July (before the full control) were used as the reference data (n = 1611) to set up a relationship between ln(VOC) and meteorological parameters. Of the reference data 70% was chosen randomly as training sample to train the neural network, 20% of the data as testing sample used to track errors during training in order to prevent overtraining, and the rest of the data (10%) as the holdout sample used to assess the final neural network for validation. The network contained one hidden layer, sigmoid functions were chosen as the activation functions for hidden layer and output layer. Given that there was no statistical difference between the prediction and the corresponding observation for validation data, the same approach was used for predicting the concentrations of VOCs from 21 July to 27 August.

2.3 Model description for OVOC simulation

A zero-dimensional box model, which is similar to the one used by Lu et al. (2012) and Li et al. (2014), was employed in this work to examine the change of OVOC production under the influence of emission restrictions. The box model uses a subset of the MCM3.2 (available online at http://mcm.leeds.ac.uk/MCM) which contains fully explicit chemical mechanisms for 57 kinds of VOCs (including 7 oxygenated hydrocarbons) plus CH4 and CO. The model calculations were constrained by measured C2–C11 NMHCs,
Aromatics, as a group of primary anthropogenic NMHCs, experienced a decrease (50–60%) in aromatics after 21 July, mainly coming from vehicle exhaust and solvent usage in Beijing (Liu et al., 2005; Yuan et al., 2010b). They dropped to values of formaldehyde, acetaldehyde, MEK and methanol before (3–20 July) and during the full control period (26 July–27 August), with 2 days spin-up time to reach steady state. The reductions of aromatics were not as notable as aromatics (Fig. 3). Some differences of ketones and aldehydes between the two time periods were found, but not as prominent as aromatics. It is difficult to assess the effectiveness of control measures on OVOC species merely based on the changes in their absolute concentrations before and during the control. In the following section a neural network analysis is used to subtract meteorological effects and quantify the influence of emission reductions on ambient levels of NMHCs and OVOCs.

3 Results and discussion based on observations

3.1 Diurnal variations of VOCs before and during the control

Table 1 summarizes the averaged meteorological parameters before (3–20 July) and during the full control period (21 July–27 August). Meteorological conditions in the two time periods were similar, except that the average ambient temperature and UV levels before the control were slightly higher than respective values during the full control, which would affect the emission of biogenic VOCs (dominated by isoprene). As an overview of the changes in VOC concentrations between the two periods, the 10 min-average diurnal variations of the mixing ratios of 12 VOC species were compared in Fig. 1. Aromatics, as a group of primary anthropogenic NMHCs, mainly come from vehicle exhaust and solvent usage in Beijing (Liu et al., 2005; Yuan et al., 2010b). They dropped to the minimum at 14:00–15:00 for both of the two time periods (Fig. 1a–f). Compared to the period prior to 20 July, an obvious decrease (50–60%) in aromatics occurred after 21 July, especially at rush hours, which may be caused by the traffic restrictions. Oxygenated species at the PKU site, from both secondary photochemical production and primary emissions, reached high concentrations at noon and in the late evening (Fig. 1h–l). Some differences of ketones and aldehydes between the two time periods were found, but not as prominent as aromatics. It is difficult to assess the effect of control measures on OVOC species merely based on the changes in their absolute concentrations before and during the control period. The following section a neural network analysis is used to subtract meteorological effects and quantify the influence of emission reductions on ambient levels of NMHCs and OVOCs.

3.2 Effect of emission control measures on mixing ratios of VOCs

Assuming that no emission restrictions had been in place from 21 July to 27 August, the “uncontrolled” VOCs corresponding to the meteorological conditions encountered at that time were predicted by the MLP network using the relationship between VOCs and meteorological parameters from the reference data (before 21 July). Figures 2 and 3 present the probability distributions of the observed (red) and predicted (blue) ln(VOC) before and during the control period for NMHCs and OVOCs, respectively. For most aromatics, the observation and prediction values from the uncontrolled period matched well with each other, but the observed data from the controlled period clearly shifted towards lower values compared to the prediction. This suggests that the influence of the traffic restriction on aromatics was effective.

Acetonitrile (CH$_3$CN), unlike other NMHCs, did not show any difference between the observed and predicted values for both the controlled and uncontrolled periods. As acetonitrile is usually considered as a tracer for biomass burning and is seldom detected in automobile emissions (Wang et al., 2007), its concentration should be hardly affected by the traffic control. The measured median value of acetonitrile before 21 July was slightly higher than that after 21 July. This discrepancy (15%) in the absolute concentrations between the two time periods could be explained by the changes in meteorological conditions or pollution transport. The reductions of most OVOCs were also observed relative to what was to be expected without emission controls, although the deviations in OVOCs were not as notable as aromatics (Fig. 3).

The predicted and observed results of NMHCs and OVOCs in the controlled period are compared in Table 2, and the results for the uncontrolled period are listed in Table S1. The median concentrations of aromatics dropped by 32–47% compared to the levels expected under the same meteorological conditions without traffic controls. The median values of formaldehyde, acetaldehyde, MEK and methanol were decreased by 12.9, 15.8, 17.1 and 19.6%, respectively. Isoprene, unlike aromatics, exhibited a broad peak from the early morning to the afternoon (Fig. 1g), which followed the...
Table 1. Summary of meteorological parameters at the PKU site before and during the full control (average ± 1σ SD).

<table>
<thead>
<tr>
<th>Time period</th>
<th>$T$ ($^\circ$)</th>
<th>RH (%)</th>
<th>Precipitation (mm)</th>
<th>WS (m s$^{-1}$)</th>
<th>UVA (W m$^{-2}$)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Before the full control</td>
<td>28.9 ± 4.5</td>
<td>66.9 ± 20.0</td>
<td>0.04 ± 0.37</td>
<td>1.1 ± 0.9</td>
<td>6.6 ± 8.7</td>
</tr>
<tr>
<td>During the full control</td>
<td>28.0 ± 4.1</td>
<td>67.9 ± 17.5</td>
<td>0.05 ± 0.51</td>
<td>0.9 ± 0.8</td>
<td>5.3 ± 7.1</td>
</tr>
</tbody>
</table>

Figure 1. The 10 min average diurnal variations of several NMHCs and OVOCs measured by PTR-MS at the PKU site before (1–20 July, blue) and during the full control (21 July–27 August, red), respectively. (a) benzene, (b) toluene, (c) styrene, (d) C8 aromatics, (e) C9 aromatics, (f) C10 aromatics, (g) isoprene, (h) MVK + MACR, (i) formaldehyde, (j) acetaldehyde, (k) acetone, (l) MEK. The error bars represent standard deviations (SDs) for those compounds before (blue) and during the full control (red).

Solar radiation and temperature cycles and showed the characteristics of local biogenic emissions in the daytime. In addition, their low concentrations at night likely indicate small local emissions from vehicles near the site. Compared to the values before the control isoprene and MVK + MACR were estimated to be reduced by 26 and 11% from 21 July to 27 August, respectively, which was caused by a combined effect of the lower temperature, solar radiation and the control measures.

The relative difference between predicted and measured acetone was as small as 1%, implying that the emission controls had little or no effect on acetone. The unchanged acetone levels were probably influenced by its high regional background, which was around 1.7–2 ppb at the PKU site accounting for 43–47% of acetone concentration (Liu et al., 2009; Yuan et al., 2012). The acetone background was closely related to transport of photochemical plumes from the surrounding areas (such as Hebei Province and Tianjin) to Beijing. As the emission reductions were implemented...
within the Beijing area, i.e., in a range of about 150 km radius, the regional background of acetone should hardly be affected by the emission controls. So, it seems to be difficult to reduce the level of acetone in Beijing if relying solely on the control of local emissions.

With the exception of acetonitrile and acetonitrile, the deviation for all other species between the full control period and the uncontrolled day is statistically significant from Student’s t test at the 95 % level (P(t)<0.05). By contrast, the reference data from the uncontrolled period were well predicted by the MLP simulations (as shown in Fig. S3), with correlation coefficients ranging from 0.79 to 0.94 and successful t tests for all the species (presented in Table S1).

Thus, the emission restrictions implemented in Beijing had significant effects on the NMHCs dominated by vehicular emissions, which is consistent with those results in Wang et al. (2010a). The reductions in OVOC species are moderate in comparison to their precursors (e.g., NMHCs), which are subject to the combined influence of controlled direct emissions, local production from reduced precursors and regional formation processes (transport). In subsequent model analyses (Sect. 4), the changes of OVOCs produced from local NMHC oxidation will be discussed.

### 3.3 Emission ratios for anthropogenic NMHCs

As NMHCs play significant roles on the formation of secondary OVOCs, emission ratios (ERs) of NMHCs before and during the full control were compared in this section to determine whether the emission restrictions also altered the relative emission of OVOC precursors (i.e., source pattern of NMHCs), in addition to reducing their total emission amounts. According to previous studies on NMHC source apportionment in Beijing (Liu et al., 2005; Song et al., 2007; Lu et al., 2007; Wang et al., 2010a), vehicle exhaust is the largest contributor to ambient NMHCs with 40–58 %, followed by solvent usage and painting processes (18–30 %), gasoline vaporization (7–13 %) and chemical plants (3–15 %).

The emission ratio for VOCs is defined as the ratio of VOC species relative to a reference compound in fresh emissions without undergoing photochemical processes. It can be determined using a photochemical age-based method presented in de Gouw et al. (2005), Warneke et al. (2007) and Borbon et al. (2013). The degradation of NMHCs by their reactions with OH is described as Eq. (1), where ER<sub>NMHC</sub> represents the emission ratio of NMHC species relative to CO, and k<sub>NMHC</sub> and k<sub>CO</sub> are the OH rate coefficients for NMHC and CO, respectively. The ERs reflect the comprehensive effect of all emission sources on ambient NMHCs at the site. Here, the photochemical age of the sampled air masses was calculated by the measured ratio of toluene/benzene (T/B) using Eq. (2), and the initial T/B was set to 2.8 ± 0.7. The 24 h averaged concentration of OH was taken as 2.5 × 10<sup>6</sup> molecules cm<sup>−3</sup> which comes from OH measurements at a suburban site near Beijing during the CAREBEIJING 2006 campaign (Lu et al., 2013). The ratio of NMHCs (e.g., toluene, m,p-xylene) to CO decreases with the photochemical age (Fig. 4). The yellow line in Fig. 4a is the linear fit between ln([toluene]/[CO]) and the photochemical age. The emission ratio of toluene relative to CO (ER<sub>toluene</sub>) is determined by the intercept of the linear fit. It is noted that a constant CO background of 120 ppbv was subtracted from measured CO when plotting VOC/CO versus photochemical age, the CO background was determined from the minimum concentrations when prevailing winds

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**Table 2.** Comparison of the observed and predicted values of median ln(VOC) and their corresponding concentrations during the full control period, and the deviations between observed and predicted concentrations, together with the outputs P(t) from Student’s t test for the two data sets, where P(t)<0.05 implies that the difference in the two data sets is statistically significant at the confidence level of 95 %.

<table>
<thead>
<tr>
<th>VOC</th>
<th>Median ln(VOC)</th>
<th>Concentration (ppbv)</th>
<th>Deviation of concentrations (%)</th>
<th>R</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>observation</td>
<td>prediction</td>
<td>P(t) observation</td>
<td>prediction</td>
</tr>
<tr>
<td>Benzene</td>
<td>−0.23</td>
<td>0.25</td>
<td>0.00</td>
<td>0.79</td>
</tr>
<tr>
<td>Toluene</td>
<td>0.24</td>
<td>0.63</td>
<td>0.00</td>
<td>1.27</td>
</tr>
<tr>
<td>C8 aromatics</td>
<td>0.12</td>
<td>0.73</td>
<td>0.00</td>
<td>1.12</td>
</tr>
<tr>
<td>C9 aromatics</td>
<td>−0.63</td>
<td>−0.17</td>
<td>0.00</td>
<td>0.53</td>
</tr>
<tr>
<td>C10 aromatics</td>
<td>−1.70</td>
<td>−1.05</td>
<td>0.00</td>
<td>0.18</td>
</tr>
<tr>
<td>Styrene</td>
<td>−2.11</td>
<td>−1.66</td>
<td>0.00</td>
<td>0.12</td>
</tr>
<tr>
<td>Acetonitrile</td>
<td>−1.45</td>
<td>−1.46</td>
<td>0.72</td>
<td>0.23</td>
</tr>
<tr>
<td>HCHO</td>
<td>1.72</td>
<td>1.85</td>
<td>0.00</td>
<td>5.56</td>
</tr>
<tr>
<td>Acetaldehyde</td>
<td>0.78</td>
<td>0.95</td>
<td>0.00</td>
<td>2.18</td>
</tr>
<tr>
<td>Acetone</td>
<td>1.39</td>
<td>1.40</td>
<td>0.36</td>
<td>4.02</td>
</tr>
<tr>
<td>MEK</td>
<td>−0.11</td>
<td>0.08</td>
<td>0.00</td>
<td>0.90</td>
</tr>
<tr>
<td>Methanol</td>
<td>1.95</td>
<td>2.17</td>
<td>0.00</td>
<td>7.06</td>
</tr>
<tr>
<td>Isoprene</td>
<td>−0.07</td>
<td>0.24</td>
<td>0.00</td>
<td>0.94</td>
</tr>
<tr>
<td>MVK + MACR</td>
<td>−0.27</td>
<td>−0.15</td>
<td>0.00</td>
<td>0.76</td>
</tr>
</tbody>
</table>
were from the northeast and relatively clean air was measured.

\[
\frac{[\text{NMHC}]}{[\text{CO}]} = \frac{[\text{NMHC}]}{[\text{CO}]} = \text{ER}_{\text{NMHC}} \\
\times \exp\left[-(k_{\text{NMHC}} - k_{\text{CO}})[\text{OH}]\Delta t\right] \\
\Delta t = \frac{1}{[\text{OH}](k_{\text{toluene}} - k_{\text{benzene}}) \times \left[\ln\left(\frac{[\text{toluene}]}{[\text{benzene}] |_{t=0}}\right)\right] \\
- \ln\left(\frac{[\text{toluene}]}{[\text{benzene}] |_{t=0}}\right) \right)
\] (2)

As discussed in Warneke et al. (2007), the error of emission ratio arises from the uncertainty of measured benzene and toluene and the estimation of the initial T/B ratio in Eq. (2). The total ER error was then calculated conservatively by linear addition of the above two errors. The measurement uncertainty of benzene and toluene by PTR-MS or GC-MS is about 10–15%. As initial T/B ratio from the same source category varies with different operating conditions, the error from selecting initial ratios can be regarded as the uncertainty of source profiles, which is usually in the range of 15–20%. So, the total uncertainty of ER estimation is around 30–35%.

Since the GC-MS/FID data was only available for the full control period of 2008, VOC data measured at the PKU site in August 2005 (Liu et al., 2009) were used as a reference for uncontrolled situations. The emission ratios for all measured hydrocarbons versus CO in Beijing for August 2008 and August 2005 are compared in Fig. 5, and the values are tabulated in Table S2 together with some ER results from two US cities in previous work (Warneke et al., 2007; Borbon et al., 2013). In general, the ERs for all groups of NMHCs in the 2 years agree with each other within a factor of 1.5 (indicated by the shaded area in Fig. 5). For traffic-related NMHCs including most alkenes, acetylene, benzene, toluene and ethylbenzene, the difference of the ERs between 2005 and 2008 ranged from ±6.5 to ±29%, within the range of combined error (30%) for ER calculations. This suggests that the pollution control measures did not significantly change the emission composition of the species closely related to vehicu-
lar emissions, although the absolute concentrations of those species did drop by half (e.g., toluene/CO in Fig. 4a). A recent work by Wang et al. (2014) also reported a good agreement of emission ratios for most NMHCs at the PKU site in the summertime of 2008, 2010 and 2011. These results reflect the similarity of NMHC compositions for typical urban emissions of Beijing.

Some alkanes (ethane, n-butane, i-butane, n-pentane) and aromatics (m,p-xylene and o-xylene, labeled in Fig. 5) showed a larger variability (35–50 %) in the two ER data sets for 2005 and 2008. For example, the emission ratio for m,p-xylene in August 2008 was observed to be 38 % lower than the value in August 2005, shown in Fig. 4b. Besides an additional error (5–10 %) from subtracting a constant CO background in ER estimation, one possible reason for this larger difference is that the abovementioned species also come from “other” emission sources in addition to vehicles. C₈ aromatics were observed as the main components from solvent usage in Beijing (Yuan et al., 2010b). The 2008 control measures included temporary closures of chemical plants, painting processes and constructions involving evaporation emissions. If the influence of the control measures on emissions of solvent usage and of automobile source occurred to different degrees, the averaged emission profiles of the city would be changed for the characteristic species (i.e., xylenes) of solvent usage.

The ERs of hydrocarbons related to vehicle emissions (acetylene, ethylene, propene, benzene and toluene) in Beijing generally agree with those in two US cities (in Table S2), indicating that emission patterns of automobile sources in different cities showed a similarity. Greater deviations of lighter alkanes, some C₄ and C₅ alkenes and higher aromatics were found probably because of the unique characteristics of emission sources (e.g., fuel types, vehicle fleet ages, solvent and paint types, industrial procedures, etc.) for each city.

4 Modeling analysis for the changes of OVOCs

To better understand the relationship between reduced precursors (NMHCs) and OVOC concentrations during the CAREBEIJING 2008 experiment, a box model using the recent version of MCM was applied in this study. According to the source apportionments of OVOCs in Beijing (Liu et al., 2009; Yuan et al., 2012), 60–70 % of aldehydes are attributed to hydrocarbon oxidation (secondary formation), the contributions from primary sources are more important for ketones (38–80 %) and alcohols (48–74 %), and the contribution of secondary formation was not detected in methanol. Therefore, the following sections will focus on the formation of aldehydes and ketones. The primary emissions for aldehydes and ketones are not included in the base model but added to subsequent model scenarios for comparison.


Table 3. Model scenarios performed in OVOC simulations during the CAREBEIJING 2008.

<table>
<thead>
<tr>
<th>Simulation</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>M0</td>
<td>MCM3.2 mechanism, $\tau_D = 24$ h, the boundary layer height (BLH) was assumed to be 1000 m, i.e., $v_{\text{depo}} = 1.2$ cm s$^{-1}$</td>
</tr>
<tr>
<td>M1</td>
<td>As M0, the variation of BLH was taken into consideration, $\tau_D = v_{\text{depo}}/\text{BLH}$, and $v_{\text{depo}}$ was set to 1.2 cm s$^{-1}$ as what was in M0</td>
</tr>
<tr>
<td>M2</td>
<td>As M1, $\tau_D = v_{\text{depo}}/\text{BLH}$, but $v_{\text{depo}}$ were different for calculated species, vertical dilution rates ($\lambda_{\text{dil, vert}}$) were added, $\lambda_{\text{dil, vert}} = \max(0, \frac{1}{\nu} \frac{dH(1)}{dt})$</td>
</tr>
<tr>
<td>M3</td>
<td>As M2, uptake of formaldehyde and acetaldehyde by aerosol were included, i.e., $\frac{dC}{dt} = -\frac{1}{2} S_{\text{RH}} \times v x C$, $\gamma$, $S_{\text{RH}}$, and $\nu$ represent uptake coefficient, RH corrected aerosol surface concentration, and molecular velocity, respectively; here we assumed $\gamma = 10^{-3}$ for HCHO and CH$_3$CHO</td>
</tr>
<tr>
<td>M4</td>
<td>As M3, using emission ratios of OVOCs relative to CO (ER$_{\text{OVOC}}$) and measured CO, the primary aldehydes and ketones were added to the model run</td>
</tr>
<tr>
<td>M5</td>
<td>As M4, the concentrations of NMHCs, CO, NO$_x$, O$_3$ were increased in order to simulate the scenario without traffic restrictions</td>
</tr>
</tbody>
</table>

(1) Vertical dilution rates were determined by the method used in Riemer et al. (2009).
(2) $S_{\text{RH}}$ is the RH-corrected aerosol surface concentration, it was estimated by an empirical function:
\[ S_{\text{RH}} = f(\text{RH}) \times S_{\text{dry}} = (1 + a \times \text{RH}) \times S_{\text{dry}}. \]
Here $a$ and $b$ were set to 4.34 and 6.72, respectively, taken from Liu et al. (2013); $S_{\text{dry}}$ was from the SMPS measurement during the campaign.

4.1 Description of model scenarios and OH simulation results

Six model scenarios for OVOC simulations are summarized in Table 3. M0 is the base case model with the settings described in Sect. 2.3. In M1, the variation of the boundary layer height (BLH) at a suburban site near Beijing taken from Liu et al. (2011) was included in the model run. The dry deposition rate in M1 was set to 1.2 cm s$^{-1}$, as used in the model scenario M0. In M2, different dry deposition rates were applied to the simulated species. The values were taken from Emmerson et al. (2007) for HNO$_3$ (2 cm s$^{-1}$), organic peroxides (0.5 cm s$^{-1}$), H$_2$O$_2$ and organic nitrates (1.1 cm s$^{-1}$), from Li et al. (2014) for HCHO and other aldehydes (1 cm s$^{-1}$). In addition, the vertical dilution rates that represent entrainment of a growing boundary layer were introduced to all modeled species in M2. These are estimated from the growth rate of boundary layer heights (Riemer et al., 2009), described by the equation given in Table 3. In M3, the loss of two aldehydes on particle surface was estimated by using the uptake coefficient of 10$^{-3}$ form the literature (Jayne et al., 1996; Li et al., 2011); the equation (Jayne et al., 1996) for calculating uptake rate is shown in Table 3. In M4, the primary sources for formaldehyde, acetaldehyde, acetone and MEK were included by employing the emission ratios of OVOCs relative to CO and measured CO concentration.

OH concentrations were not directly measured in August 2008. As strong relations between observed OH and ozone photolysis frequency ($J(O^1D)$) were found in field stud-

ies (Rohrer et al., 2014; Rohrer and Berresheim, 2006), it makes sense to estimate OH radicals using the empirical OH-$J(O^1D)$ relations when lacking direct measurements of OH. Here, the modeled OH by the box model was compared with the calculated OH by an observed dependence of OH on $J(O^1D)$ that was obtained from OH measurements in CAREBEIJING 2006 (Lu et al., 2013). Figure 6a shows the average diurnal profiles of modeled and calculated OH, indicating that the modeled OH in the daytime shows a broader peak and it was overestimated by 17% in the calculations from the empirical relation. Relatively higher nighttime OH was determined from the calculation, due to the intercept of the empirical OH-$J(D^1D)$ function. This intercept includes all processes that are light independent (Rohrer and Berresheim, 2006), implying that unknown OH production probably occurred at night (Lu et al., 2014). As the uncertainties of OH measurement and $J(O^1D)$ measurement are 20 and 10%, respectively, from Lu et al. (2013), in addition to a fitting error between OH and $J(O^1D)$, the difference between the modeled and calculated OH is acceptable. This hints that the box model employed here provides a reasonable explanation of radical chemistry in the urban environments.

4.2 OVOCs simulation

4.2.1 HCHO and CH$_3$ CHO

Unlike OH, a large difference with a factor of 2–4 was found between the measured and modeled concentrations of aldehydes by the base model (M0), as summarized in Table S3.
After modifying deposition rates or/and vertical dilution, the averaged concentrations of HCHO and acetaldehyde in M1 and M2 were reduced by 30–37% compared to the base case. But the modeled concentrations of the two aldehydes are still higher (>1.5 times) than the measured values. The considerations of BLH variation, different deposition and vertical dilution can help to explain 40–50% of the large discrepancy between M0 and observations. These results are consistent with a recent study on modeling of formaldehyde and glyoxal in the Pearl River Delta (PRD) (Li et al., 2014).

Li et al. (2014) estimated that the uptake of formaldehyde and glyoxal by aerosols had the largest contribution (∼50%) to aldehyde sinks in the presence of acidic aerosols. The $H_{\text{aer}}^+$ presented in the particle phase was calculated from inorganic ions (NH$_4^+$, SO$_4^{2-}$, NO$_3^-$ and Cl$^-$) by HR-TOF-AMS using the method in Zhang et al. (2007). For most days of campaign the average acidity of aerosols was close to neutral due to high ammonia in Beijing. However, for several days (such as 31 July, 1, 4, 11 and 14 August) the averaged $H_{\text{aer}}^+$ in the daytime was up to 0.01448 mol L$^{-1}$, corresponding to a pH value of 1.84, which indicates high aerosol acidity occasionally occurred during the campaign. Jayne et al. (1996) reported a large uptake of formaldehyde by aqueous surface at low temperature and high aerosol acidity. Some lab experiments (Li et al., 2011) and field studies also showed loss of HCHO on aerosols is possible and driven by the liquid water content of the aerosol phase (Toda et al., 2014). In this study, the loss of aldehydes on aerosol particles through heterogeneous uptake processes was included in M3 and M4 by using the uptake coefficient of 10$^{-3}$ for two aldehydes. On average, the modeled HCHO and CH$_3$CHO by M3 were decreased by 64 and 58%, respectively, compared to M2. Therefore, the loss of aldehydes on aerosol particles might be important in the polluted areas with high production rates of aerosols. Further research on sinks of aldehydes, particularly for heterogeneous uptake processes, is still needed in future studies.

The model scenarios M0–M3 only estimated photochemical production of OVOCs (secondary OVOCs) and did not include their primary emissions from sources. As OVOC source apportionment showed that Beijing is characterized by high anthropogenic emissions of OVOCs (primary OVOCs) (Yuan et al., 2012; Chen et al., 2014), primary aldehydes were added to model simulation M4 by using the ERs of OVOCs relative to CO (HCHO/CO = 2.74, CH$_3$CHO/CO = 2.82) and measured CO. The ERs of OVOCs were calculated from the multivariate regression results for OVOC data at the PKU site in Yuan et al. (2012). Compared with M0–M3, the average of modeled HCHO and CH$_3$CHO by M4 show a better agreement with the observations. As shown in Fig. 6b and c, the maximum of modeled HCHO in M4 occurred at 15:00–16:00 LT, which was delayed by about 3 h from the observations. The diurnal pattern of simulated acetaldehyde was well matched with the observed variation most of the time. The calculated primary HCHO and CH$_3$CHO by M4 contributed 41 and 58% of the total modeled concentrations, respectively, higher than the correspond-
ing results (33% for HCHO and 42% for CH$_3$CHO) from the PMF model (Chen et al., 2014).

4.2.2 MACR

The model overpredicted the overall concentrations of MVK + MACR by 25% (Table S3). The diurnal cycle of modeled MVK + MACR is similar in shape and magnitude to the measurements for early morning and nighttime, while the model overestimated the midday peak by a factor of 1.5 in the afternoon (Fig. 6d). This suggests that the box model is able to represent nocturnal sinks of MVK + MACR, but cannot fully explain their productions or/and sinks in the daytime. Karl et al. (2010) found that the observed deposition velocities for MVK + MACR in tropical ecosystems were up to 2.4 cm s$^{-1}$, 3–4 times higher than as used in our model runs (0.6–0.8 cm s$^{-1}$). They also reported that the uptake of MVK + MACR by vegetation followed an exponential increase with leaf temperature, and presented a light dependency as well. So, more MVK + MACR possibly deposits on leaves in the afternoon, which gives evidence of the gap between modeled and observed MVK + MACR during that period.

4.2.3 Acetone and MEK

The modeled concentrations of MEK in M0 underestimated the measurements by 56% (given in Table S3), which is probably caused by the absence of direct anthropogenic emissions of MEK in the model. Sommariva et al. (2011) also found the underprediction of ketones using the box model in the ICARTT (International Consortium for Atmospheric Research on Transport and Transformation) campaign. Similar to HCHO and CH$_3$CHO, the primary ketones were added in M4 by using measured CO and emission ratios of ketones (acetone / CO = 2.23, MEK / CO = 1.22). The modeled ketones in M4 were comparable to the measurements, as listed in Table S3. For diurnal variations, the agreement between modeled and measured MEK concentrations is generally good (shown in Fig. 6f). Figure 7 indicates the simulated contributions of primary and secondary sources to MEK at the PKU site. Primary sources dominated the ambient MEK concentration (~80%) on a 24 h basis, and the contribution of secondary MEK increased to 25% at noon and early afternoon.

The model cannot reproduce the observed variations of acetone. The acetone peak at noon was probably attributed to some temperature-related emissions (e.g., solvent evaporation). As direct emissions of acetone have not always kept pace with CO, the primary acetone derived from the ER expressed by the acetone / CO ratio cannot explain it. Besides combustion sources, in urban areas acetone also directly comes from solvent usage and evaporation, particularly in chemical processing procedures. So, it needs to find a slightly reactive or inert species to work as an indicator of non-combustion sources of acetone. CO and acetylene are frequently used as reference compounds in ERs, while they come mostly from automobile exhaust and fuel combustion, they were seldom observed in solvent usage. Other relatively inert VOCs such as ethane and propane show more variable emission ratios for different sources, so they are also not suitable to be a unique tracer for non-combustion sources. Thus, more detailed measurements on emission characteristics of acetone sources need to be investigated in further studies.

In summary, the box model overestimated the concentrations of aldehydes by a factor of 1.4–1.7 without considering the uptake by aerosols. This discrepancy is mainly attributed to missing sinks, such as vertical dilution, transport and heterogeneous uptake on aerosols. In the presence of fresh emissions, the box model predicted the concentrations and variations of MEK well, but it cannot explain observed acetone and is likely to be affected by unidentified primary emissions or high backgrounds. Despite the existence of drawbacks on estimation of sinks, the box model can still give an overview of the relative importance of different precursors on the production of secondary OVOCs. It can also be used as a tool to quantify the relative changes in OVOC production due to the reduced precursors and other pollutants during the emission control period.

4.3 Secondary production of OVOCs

Figure 8a–c illustrate averaged diurnal variations of the production of formaldehyde, acetaldehyde and acetone from different groups of NMHCs in M2 during the full control period. As an average, the photooxidation of alkenes contributed most to secondary HCHO production with a percentage of 54%, followed by the oxidation of isoprene (30%) and aromatics (15%), the contributions of pinenes and alkanes were negligible (<1%). The contributions of C$_2$–C$_4$
alkenes and isoprene were well matched, accounting for 39.7 and 37.5% of the total of averaged HCHO production, respectively (listed in Table S4).

Acetaldehyde production was found to be dominated by alkene oxidation (91%) and the contributions of alkanes and aromatics were minor (7 and 1%, respectively). Usually the dominant secondary CH$_3$CHO source is from alkane oxidation (ethane, n-butane, i-butane) via the reaction of the C2H5O2 peroxy radical (Carter, 1990; Sommariva et al., 2008). But Sommariva et al. (2011) found that in urban plumes in the northeastern US, propene and other alkenes significantly contributed to the formation of acetaldehyde via the reaction of the HYPROPO alkoxy radical, especially in more fresh air masses during the first 2 days, when their contribution to acetaldehyde production is more important (up to 25%). Then the role of C3–C5 alkenes decreases very quickly because of their high reactivity. Compared to urban plumes in the USA, air masses in Beijing are much fresher and closer to the emission sources. In Fig. 4 we can see that the majority of photochemical ages of air masses during the campaign are shorter than 40 h, i.e., within 2 days. In this case, alkene oxidations become the most important processes for CH3CHO formation.

For acetone, 60% of its production was from alkenes, 29% from alkanes, 10% from pinenes, and the remaining 1% was attributed to aromatics and isoprene. Secondary MEK was predominantly produced from the oxidation of alkanes (>95%). The averaged results show that the oxidation of anthropogenic precursors is the dominant production pathway for secondary OVOCs in Beijing.

The NMHC oxidation for all OVOC compounds displays an obvious diurnal pattern with a peak at noon. Table 4 summarizes the averaged production rates and corresponding percentages of HCHO from different NMHC groups. For the full control period (M4), the pathway of isoprene oxidation accounted for up to 40% of the total production in the daytime, comparable to that of anthropogenic alkenes (46%). The remaining (13%) of HCHO production was attributed to aromatics. The contribution of biogenic sources decreased to 23% during the night and, accordingly, anthropogenic sources became the dominant contributor to HCHO production with a combined percentage of 76% (alkenes + aromatics). The differences in contributions between day and night are closely related to different diurnal patterns of anthropogenic and biogenic NMHCs.

### 4.4 Changes in OVOCs responding to the control measures

As indicated above, ambient concentrations of NMHCs were reduced by 30–45% due to the decrease in the city’s emissions from 21 July to 27 August. Here, the effect of the reduced precursors together with other gas pollutants (CO, NO, NO$_2$ and O$_3$) on secondary OVOCs has been estimated using the box model.

Since the whole group of NMHCs were not measured by online GC-MS before the full control (3–20 July), the model scenario M5 ran with increased concentrations of alkanes, alkenes, aromatics and isoprene by 30, 35, 40 and 15%, respectively, according to the results from this work and a previous study on NMHCs by Wang et al. (2010a), in order to simulate the situation of the uncontrolled period. The concentrations of NO, NO$_2$ and CO in M5 were increased by 35, 25 and 17%, while O$_3$ was decreased by 16%, taken from the observed changes in Wang et al. (2010c) and Chou et al. (2011). Then, the relative changes of secondary OVOCs responding to the control measures were determined by the difference between the model scenario of M5 (the uncontrolled period) and M4 (the full control period).
The modeled reductions in HCHO, CH₂CHO and MVK + MACR from M5 to M4 agree quite well with the corresponding results from the MLP network (given in Table 5), suggesting the measurable changes of those three species can be reasonably represented by the photochemical production from the oxidation of reduced precursors in the box model. For MEK, the relative change from the box model is 3.7% lower than the MLP results, which is likely related to the uncertainty of primary emissions. However, we found a large discrepancy of acetone changes in the two methods. The modeled changes (12.8%) of acetone were much larger than the MLP result (1%). As discussed in Sect. 3.2, there is no significant change in measured acetone during the full control period probably due to high backgrounds in the region. The change (12%) of secondary production of acetone estimated by the model is relatively less important compared to transport in Beijing, it was covered by the high levels of acetone from a neighboring region. It seems difficult for the box model to evaluate such effects of background or transport.

The HCHO production rates from most of the precursors have been lowered during the full control period, but the ranks of different precursors did not change much (details listed in Table S4). Due to the emission restrictions the total production rate of HCHO decreased from 7.50 to 6.75 ppb h⁻¹, and the daytime productions of HCHO from the oxidation of alkenes, aromatics and isoprene were reduced by 13, 12 and 6%, respectively (from Table 4). The relative contribution of alkene oxidations in the daytime was decreased by 1.5% from the uncontrolled days (M5) to the full control period (M4) and, accordingly, the contribution of isoprene was increased by 1.7%. Here, it should be noted that the relative importance of isoprene chemistry has been enhanced when the NMHC precursors from anthropogenic sources were reduced by 30–40%. This would affect the relative potentials of reactive NMHCs for the formation of secondary pollutants (such as OVOCs and ozone).

5 Conclusions

In the summer of 2008 the pollution control measures implemented in Beijing provide a unique opportunity for studying how the primary and secondary air pollutants react to the reductions of anthropogenic emissions. Using a neural network analysis, we concluded that the emission controls taken in Beijing had a notable effect on reducing the ambient concentrations of formaldehyde, acetaldehyde, MEK and methanol. The influence of emission restrictions on OVOCs (except acetone) was found to be statistically significant when the variations due to meteorological conditions were excluded by the MLP network analysis. However, the effects of local emission controls on acetone seem to have been submerged in high background levels and no obvious changes were detected in acetone during the full control period. This highlights the complexity of secondary air pollutants (like...
The chemical production and degradation of OVOCs in Beijing under the emission control conditions have been determined quantitatively using a box model constrained with measurements of OVOCs precursors, gas pollutants and other physical parameters. The most important precursor of HCHO is isoprene, accounting for 30% of the total HCHO production. However, as a whole, the anthropogenic source is still the main contributor to secondary HCHO formation, with alkene oxidation accounting for 54% and aromatics for 15%. Unlike HCHO, anthropogenic NMHCs were found to play predominant roles on the chemical production of acetaldehyde, acetone and MEK. Approximately 90% of acetaldehyde production can be attributed to the oxidation of alkene. Anthropogenic alkenes and alkane contributed 60 and 29% of acetone formation, respectively, and biogenic precursors (pinenes) contributed the other 10%. The relative changes in the modeled aldehydes and MVK+MACR due to the emission restrictions are shown to be consistent with the estimated results from the neural network, reflecting that the current chemical mechanisms can largely represent the realistic formation processes of aldehydes and MVK+MACR in the high NOx urban conditions.

Compared to measurements of OVOCs, the box model overestimated the measured aldehydes by a factor of 1.4–1.7 without taking into consideration their loss on aerosols, and it was able to roughly explain diurnal variations of acetaldehyde most of time. The model–measurement discrepancy for aldehydes is mostly caused by missing sinks in the model, including physical dilution, transport and heterogeneous processes on the surface of aerosols. After adding the primary concentrations of ketones to the box model, the concentrations and diurnal cycles of MEK can be well represented by the model simulations. But the modeled changes of acetone become unimportant due to the high background and transport of acetone in Beijing and its surrounding areas.

The analysis of the summer 2008 situation in Beijing improves our understanding of the complicated relationship between anthropogenic emissions and pollution levels of primary and secondary VOCs. The concentrations of NMHCs were reduced by 30–40% due to the emission controls. As vehicle emissions are the most important sources of VOCs in the city, the emission ratios of the traffic-related species (C2 and C3 alkenes and light aromatics) in the controlled days are found to be fairly similar to the values before the controls and in other urban areas. The relatively similar emission ratios for the above hydrocarbons in cities will help in providing a general idea of the photochemical evolution of gas-phase organic carbons in urban plumes based on model simulations. As China is currently in the midst of a sustained effort to improve the air quality in its megacities, the results presented in this study can serve as a case for further understanding of the atmospheric chemistry not only for Beijing but also for large regions and as an important aid for policymakers considering ways to reduce pollution more efficiently in the long term.

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