Referee 1

The author conducted a one-year VOCs observation in Beijing, attempting to illustrate NO3 chemistry, especially the oxidation of VOCs by NO3. Then, a parameterization method was established to attempt to construct the long-term reactivity of NO3 using VOCs data. It was found that since 2011, the NO3 reactivity of VOCs in Beijing has been decreasing year by year, and there is a significant correlation with VOCs emission reduction. The overall research objectives of the article are clear, the research methods are appropriate, and the research conclusions are relatively reliable. However, there are still issues that need to be clarified, as follows.

Thanks for the review's constructive comments.

Major comments:

1. Correlation diagram, meaning of horizontal and vertical coordinates. For example, Figure S1. Why use simulated values as the abscissa? Generally, the abscissa is the reference value, or the true value. In this case, it is obvious that observation values should be used as the abscissa. Please review the entire article by the author.

    We have reviewed the entire article and revised all figures containing simulated and actual values. In Figure S1 (Figure S2 now), we used the observed concentration of isoprene as the abscissa and the concentration of isoprene simulated by WRF/CMAQ as the ordinate to more clearly demonstrate the relationship between the simulated and observed values. Since the reactivity containing monoterpenes was estimated, in Figure S8 and Figure S10 of section 2.2, we used the reactivity as the ordinate and used VOC concentration as the abscissa, which can better demonstrate the relationship between VOC concentration and NO$_3$ reactivity towards VOCs.

    In brief, we revised the figures as follows (Figure S2 for example).
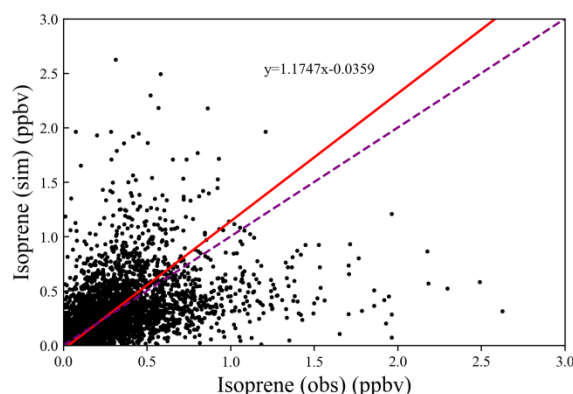

Figure S2. The intercomparing of measured and simulated isoprene in year 2019 in urban Beijing, which shows the overall overestimation of the modelling result compared with the observation.

2. Why is temperature and humidity linear between April and May? Is there no observed data? If not, please delete it. The uptake coefficient of N2O5 also has this issue.

Yes. We have removed the "linear" section because there were no values in the observation data of temperature and relative humidity from April to May. Due to the fitting relationship between the uptake coefficient with temperature and relative humidity in the heterogeneous reaction of N_2O_5, the uptake coefficient of N_2O_5 had no values from April to May, so the steady-state concentration of NO_3 also had no values.

We have revised Figure S2 (now Figure S3) as follows
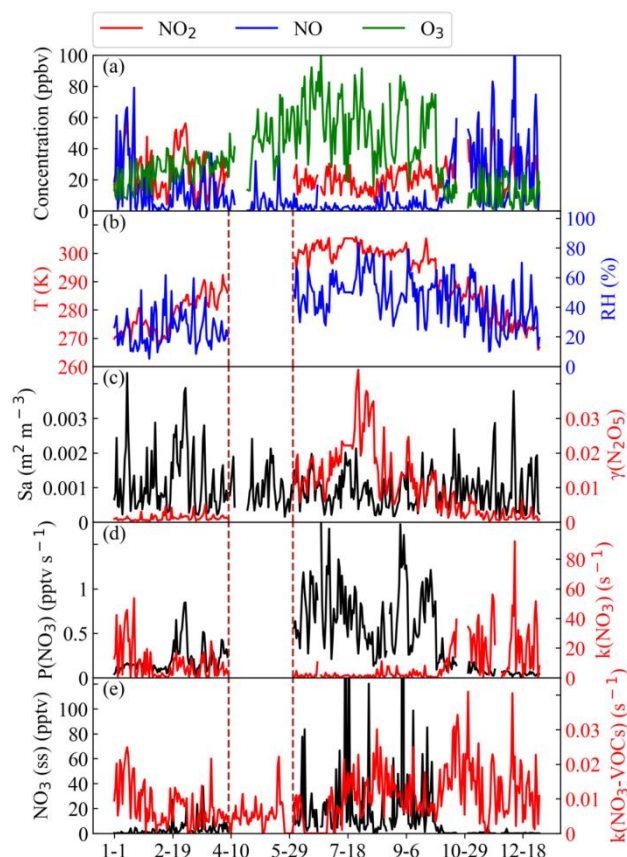
3. The red lines of d and e in Figure S2 are both NO3 losses, with the former being the total loss and the latter being the loss in reaction with VOCs. It seems that there is a significant difference between the two orders of magnitude, is it due to the participation of heterogeneous processes?

It isn't due to the participation of heterogeneous processes. In our study, we found that the concentration of NO in Beijing were relatively high, which led to NO_3 mainly lost through the reaction with NO (which can be clearly seen in the fitting diagram of R_{NO3} and NO concentration in Section 3.5 (Section 3.4 now)). Therefore, the significant difference between these two orders of magnitude is due to the higher concentrations of NO which dominated the loss of NO_3. We added the explanation of the loss of NO_3 in the caption as follows.

Figure S3. (a) Time series of concentrations of NO, NO_2 and O_3. (b) Time series of thermodynamic temperature and relative humidity. (c) Time series of the aerosol surface area densities and dimensionless uptake coefficients of N_2O_5, (d) Time series of NO_3 production rate and NO_3 loss term (the reaction with NO dominated the reactivity). (e) Time series of NO_3 stationary-state concentrations and the NO_3 loss term through the reaction with VOC (actual NO_3 reactivity towards VOC). All the data was averaged with the time resolution of 1 day.

4. Figure S7 shows that current parameterization methods can capture changes in monthly and daily averages, but cannot capture changes in hourly averages. Why? After adding terpenes in Figure S10, the fitting results actually look worse. Why?

Thanks for the interesting comment. The parameterization scheme selected the indicator species on a monthly scale, and the derivation of parameterization used the hourly data set. Therefore, the fitting result should be best in the monthly and daily scale since the goal of the parameterization is reproduced the result on a monthly trend. We confirmed that the estimated averaged diel variations of reactivity cannot capture the changes in the highest time resolution with hourly averages. However, due to the relatively stable indicating effect of VOC every month, the estimated averaged diel variations of reactivity based on months are also consistent with the actual averaged level and magnitude.

After adding the contribution of monoterpenes, the errors from the simulated concentration of monoterpenes were introduced in the actual reactivity, and there were also certain errors in parameterized estimation. The two types of errors together caused larger uncertainty in the fitting results in Figure S10 (Figure S11 now), but the averaged level and magnitude of the fitting results are relatively consistent. In future studies, the observation data of monoterpene will help to further update and improve the performance of these parameterization schemes.

5. The article discusses the NO3 reactivity of VOCs, but there is no VOCs concentration sequence diagram in the article, especially the proportion of different components, seasonal changes, daily changes, etc. In addition, the author needs to explain the detection method of VOCs and what are the pre freezing stages? Detect which species and so on, for example, indicate 56 PAMS species.

Thanks for the comments. We depicted the averaged-diel profiles, averaged-seasonal profiles and time series of concentrations of alkanes, alkenes, aromatics, and isoprene, styrene, cis-2-butene, trans-2-butene, and trans-2-pentene, and added them in the supporting information (as Figure S1 in the revised version). We added an explanation of detection method of VOC and pre-freezing stages, and classified and numbered the 56 observed VOCs in Table S1.

Line 124. The volatile organic compounds were pretreated by pre-freezing and collected in the deactivated quartz empty capillary at extreme-low temperature (-150 ℃), then heated and

delivered into the analysis system. After separation by the double chromatographic column, the low carbon compounds C2-C4 were detected by the FID detector, and the high carbon compounds C5-C10 were detected by the MS detector. There are 56 kinds of VOC are measured in total and accounted in this study (listed in Table S1 and the concentrations are depicted in Figure S1), in which monoterpenes measurement are not valid.
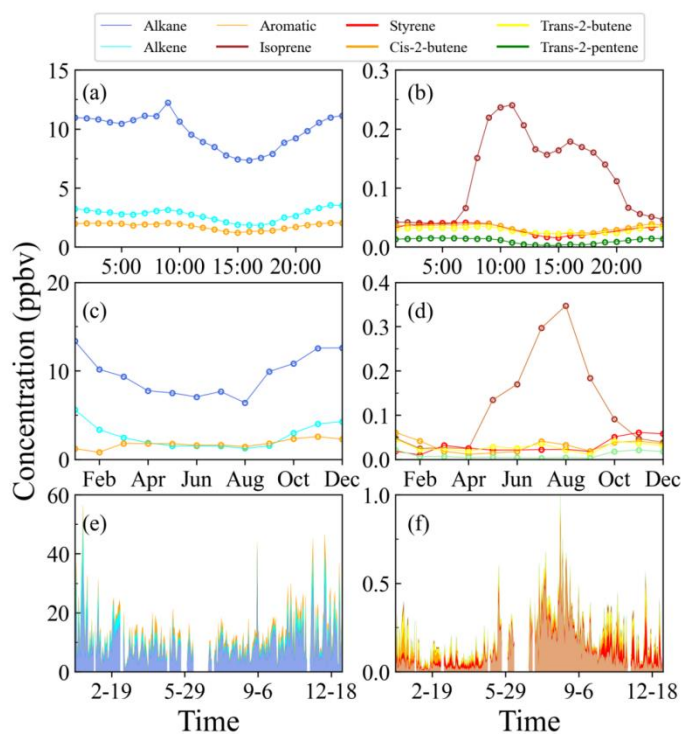
Please find the Table S1 in the updated SI.

Figure S1. The yearly-averaged diel profiles (a), monthly-averaged profiles (c) and time series (e) of concentrations of alkanes, alkenes and aromatics. The yearly-averaged diel profiles (b), monthly-averaged profiles (d) and time series (f) of concentrations of isoprene, styrene, cis-2-butene, trans-2-butene and trans-2-pentene. The colors indicate the different VOC as they are donated in the legend.

6. The article emphasizes that the simulation results of terpenes come from WRF-CMAQ. What specific simulation scheme is used? Given the important contribution of terpenes, it is recommended that the author provide a detailed explanation.

We agreed with this suggestion and added more description about the model simulation setup in the revised manuscript as follows.

Line143. The regional model CMAQ (Community Multiscale Air Quality) version 5.2 was applied to simulate air quality in the eastern China, with a horizontal resolution of 36 km. Specifically, the gas-phase mechanism of SAPRC-07 and aerosol module AERO6 were used. The meteorological fields were provided by Weather Research & Forecasting (WRF) Model version 4.2. The biogenic emissions were simulated by the MEGANv2.1, which was driven by WRF as well, and the emissions of open burning were estimated with FINN. The MEIC

7. The author used the ratio of isoprene to terpene to calculate the concentration of terpene, which may require supporting evidence. Firstly, the emission of terpenes from biological sources is related to temperature, while the emission of isoprene from biological sources is related to temperature and radiation. There may be a relationship between the two at night, but the relationship may not be significant during the day. Secondly, the author also emphasizes that motor vehicles emit a large amount of isoprene, and if the ratio is used to determine terpenes, it may lead to an overestimation of terpenes.

We highly appreciate the suggestion. As you mentioned, the factors affecting the concentrations of isoprene and monoterpene are different, so we did not use the same parameters to simulate the concentrations of the two. Instead, we used the model to simulate the relationship (or the ratio) between the two to reflect this systematic difference. To be honest, we propose that this method is reasonable and feasible while there is no more evidence since long term measurement of both isoprene and monoterpene is very scarce. This is the best what we can do to estimate the contribution of terpene. We hope to provide more evidence to prove this relationship in future research.

The isoprene emissions from motor vehicles are important parts of anthropogenic isoprene emissions in Beijing, but the anthropogenic isoprene emissions can be ignored compared to biological sources. Many studies have shown that anthropogenic isoprene emissions are less important in Beijing: The isoprene emissions of biological sources in Beijing were one order of magnetic larger than that of anthropogenic sources (Yuan et al., 2009). This indicates the concentration of isoprene at the environmental level in the urban areas of Beijing is not affected by the traffic vehicles, but mainly by plants in Beijing (Cheng et al., 2018). Therefore, the anthropogenic isoprene emissions can be ignored and the concentrations obtained by model were able to estimate without significant overestimation in our study.

Otherwise, since the previous introduction of isoprene emissions from motor vehicles in the article may provide readers with a confused understanding of isoprene in Beijing, we have deleted it in the paragraph.

8. Since WRF-CMAQ can simulate OH concentration, why use J1D to calculate OH concentration?

Here the parameterization based on the JO^1D data is based on the relationship derived from the field observation. Actually, we cannot say which OH dataset would be better to reflect the real OH level in the atmosphere. Therefore, we compared the WRF-CMAQ modelled OH

and the JO^{1}D parameterized results. As shown in the following figure, the inter-comparison confirmed that these two methods had a similar and consistent performance with a controllable difference (although the model simulation is systemically higher than fitted result with 41%). We tested the usage of modelled OH in the calculation as shown in the following analysis, although the CMAQ modelled OH enlarged the daytime oxidation, it doesn't have a significant impact on the results of VOC nocturnal oxidation. Therefore, we did not change the use of JO^{1}D parameterization in the revised manuscript.
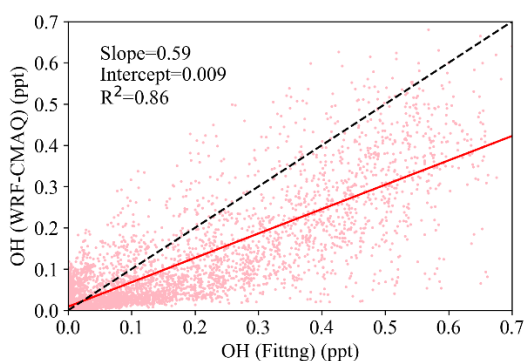
9. Isn't Equation 11 the method for calculating NO3 reactivity? What is the difference between this equation and equation 1? It seems that the difference is only slightly fewer species.

Yes. Parameterization scheme 2 is equivalent to calculate the addition of NO$_3$ reactivity towards important contributing VOC. For the Beijing region, it is the addition of NO$_3$ reactivity towards isoprene, styrene, cis-2-butene, trans-2-butene, trans-2-pentene and propylene. Since not all observed VOCs were calculated, it is called the second type of parameterization method. The significance of this method is to greatly reduce the requirements for VOC observation, which ignores VOC species with small contributions to obtain NO$_3$ reactivity towards VOC.

10. Since terpenes have made significant contributions, I think it is meaningless without adding the explanation of terpenes in sections 3.1 and 3.2. I suggest deleting section 3.3 and merging it into 3.1 and 3.2 for discussion and explanation. The author can use the simulation results of long-term terpenes to illustrate the interannual trend of NO3 reactivity of VOCs in Beijing.

We added the monoterpene reactivity and its parameterization scheme in section 3.3 to the corresponding part of sections 3.1 and 3.2, respectively.

The simulation results of long-term terpenes are lacking, but we speculate that the changes of long-term concentrations of BVOC such as terpenes may not be significant as you suggested that it is close related to the temperature, therefore the interannual change of NO$_3$ reactivity towards VOC may not be clearly demonstrated using long-term modeled terpene concentrations.

The NO_3 reactivity towards MNTs (named as k_{NO3_MNTs}) was estimated by the method mentioned in section 2.2. After taking MNTs into account, the total k_{NO3} (named as k_{NO3_total}) was greatly enlarged, with campaign-averaged value of 0.0061 ± 0.0088 s^{-1}, resulting in our results comparable with previous research results. The NO_3 reactivity towards MNTs was higher in autumn and winter and lower in spring and summer (Fig. S7). Considering the corresponding reactivity towards monoterpenes, the total NO_3 reactivity towards VOC changed from (summer > autumn > winter > spring) to (autumn > winter > summer > spring), highlighting the impact of the monoterpene variations on the reactivity. The NO_3 reactivity towards MNTs displayed significant differences between daytime and nighttime (Fig. S7c-d). The reactivity at night in all months was higher than that in the daytime, especially from October to January, highlights the role of biogenic monoterpenes in nocturnal NO_3 chemistry (Li et al., 2013; Riba et al., 1987). To evaluate the contribution of monoterpenes to the total k_{NO3}, we calculated the fraction (FMNTs) by Eq. 12.

$$F_{MNTs} = \frac{k_{NO_3\_MNTs}}{k_{NO_3\_total}} \qquad \text{Eq. 12}$$

Figure 2a displays the differences between the k_{NO3_mea} and k_{NO3_total}. Monoterpenes were very important for NO_3 reactivity, and the F_{MNTs} varied from 40% to 80%, with strong seasonal variations. The MNTs accounted for NO_3 reactivity nearly 80% in winter and spring. In the seasons when isoprene no longer dominated, the measured reactivity accounted for a small fraction, and the corresponding reactivity towards AVOC such as styrene was smaller than that of monoterpenes. As shown in Fig. 2b, the measured VOC had high fractions in the daytime and low at night. Especially in May and August. The measured VOC in the daytime accounted for more than 60% of k_{NO3_total}, which was closely related to the increasing concentrations of isoprene in the summer daytime. The reactivity towards MNTs accounted for a large fraction of reactivity at night.
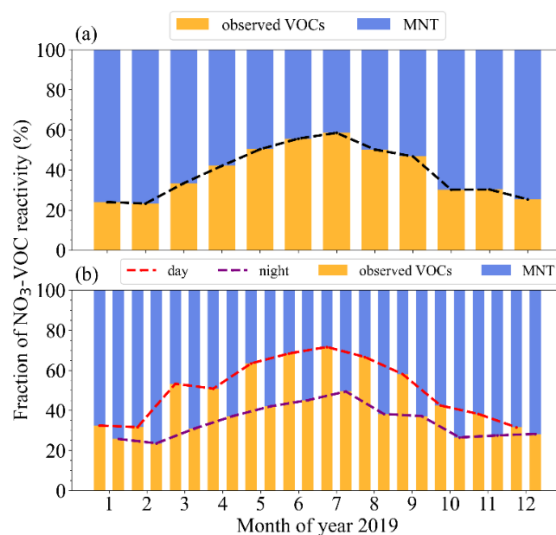
Figure 2. (a) Fractions of the k_{NO3_total}. (b) Fractions of the k_{NO3_total} divided into daytime (left) and nighttime (right). The colors on the stacked bar plot indicate the different fractions as they are donated in the legend. The lines represent the monthly-averaged variations of the NO_3 reactivity towards MNTs.

Line 346 (Section 3.2)

After taking MNTs into account, we updated the parameterization method established before by using the relationship between reactivity and VOC concentrations including monoterpenes. The updated parameterization Method 1 used the same principle as introduced in Sect 3.2, with fitting slopes changing significantly (Figure. S10). Table S5 gives the specific correlation coefficients between six key VOC concentrations and k_{NO3_total}. The updated Method 2 considered the sum contributions of six VOC and the estimated MNTs by isoprene concentration. We reevaluated the two updated parameterization methods (single VOC and six VOC, respectively). Overall, the performance of two methods are reasonable and the updated Method 1 is better than that of Method 2 in general (Fig. S11). For robustness, we evaluated this parameterization on datasets of other years (shown in Fig S12).