Point-to-point response

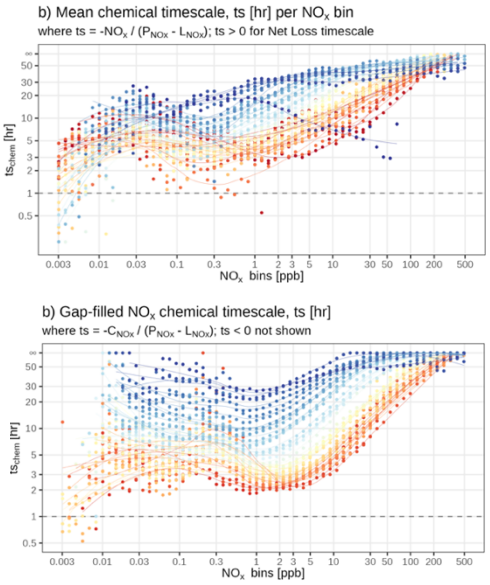
We thank two reviewers for their constructive comments and have addressed all comments with additional analyses and clarifications to the manuscript. We start the response with a summary of changes for both reviewers. Our point-to-point responses are highlighted in blue with track changes in blue/red and reviewers' comments in black.

General responses to both reviewers

Two main criticisms from the two reviewers include:

1) Omission of chemical pathways and processes (heterogeneous NOx chemistry, NO + HO2/RO2);
2) Lagrangian atmospheric mixing (mixing scales in PBL, mixing between FT and ML).

We also identified a minor bug in the estimated NOx net timescales and reran simulations. The main difference from the initial submission (**upper right panel**) is the timescale over NOx-limited regimes at high SZA bins (NOx < 0.1 ppb and SZA > 50 degrees in the **lower right panel**).



As changes in the TROPOMI version and model parameters may affect model-data comparisons, we summarize the model-data slopes for all overpasses over the New Madrid power plant based on each configuration (**Fig. R1 below**). All simulations presented here used EPA emissions and 3km HRRR meteorological fields. Similar to **Fig. 6**, not accounting for NOx chemistry results in a positive model bias of tNO2 (**blue crosses in Fig. R1**); and using the latest TROPOMI version (v2.3 from v1.3) has largely reduced the model-data mismatches (**orange circles to orange dots**). The minor correction in the derived net loss timescale and the mixing time scale (from 3 to 1hr) only slightly alter the model-data slopes (**orange dots vs. purple dots vs. purple crosses**).

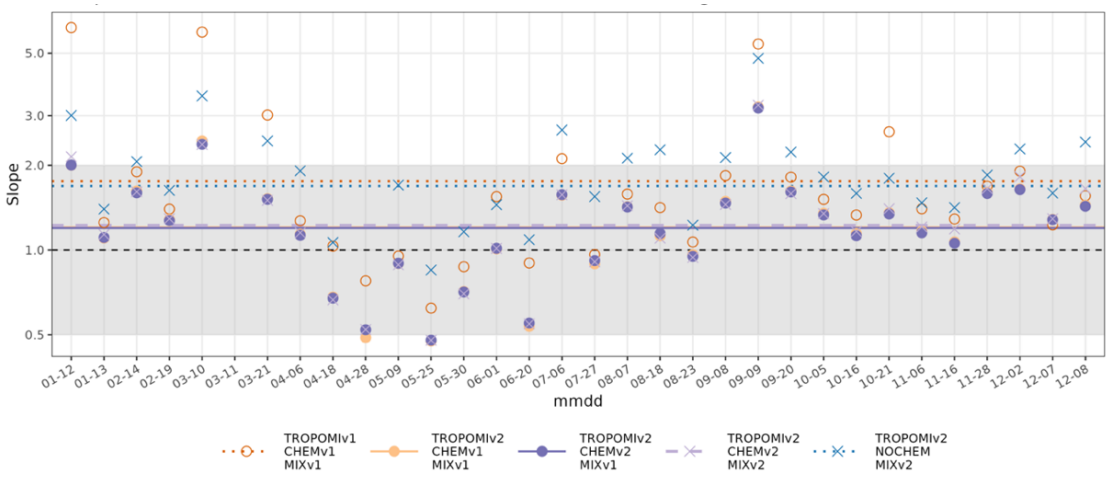


Figure R1. Model-data comparisons informed by the linear slope between the two using different versions (v1/v2) are explained as follows: **TROPOMI**: v1.3 or v2.3 of the TROPOMI L2 NO_2 retrieval. **CHEM**: v1 vs. v2 for runs using NOx curves from the initial vs. revised manuscript (as shown in the above comparison). **MIX**: inter-parcel mixing with two different horizontal mixing timescales tested (3 vs. 1 hr) over a 1km box. Tests with a spectrum of mixing parameters have been conducted in response to the 1st comment of reviewer 2.

Comments from reviewer #1

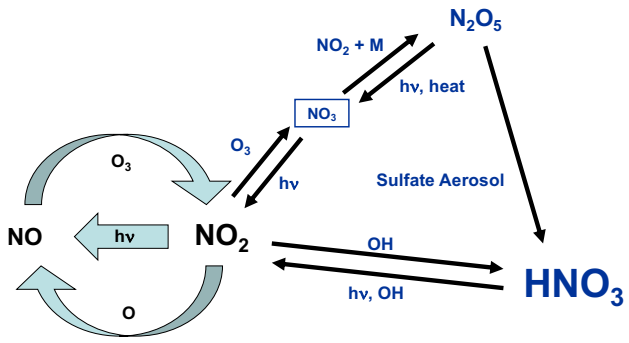
Satellite retrievals of NO2 columns are used to determine NOx emissions from power plants and cities. They are increasingly used alongside CO2 retrievals to calculate emissions from these sources. However, the effect of NOx chemistry and transport on NO_2 columns is often overlooked. To address this issue, Wu et al. have developed a model that incorporates a simplified representation of NOx chemical loss within the STILT Lagrangian particle dispersion model. It includes additional features such as a column weighting module to account for retrieval averaging kernel profiles and an error analysis module. The model is evaluated against TROPOMI NO2 observations from three power plants and two cities. The manuscript covers the model's advantages, limitations, and applications such as using NO2-to-CO2 enhancement ratios to estimate CO2 emissions and identifying wind biases in meteorological data.

While this work is generally sound and well-presented, there are areas that I think need attention.

We appreciate the constructive comments from Reviewer #1 and tried to address all concerns via additional supporting analyses.

NOx chemical tendency (Sect. 2.1):

1. It appears that the model excludes heterogeneous NOx chemistry in aerosols. If this is indeed the case, it is important to discuss the resulting errors arising from this omission. Alternatively, if heterogeneous NOx chemistry is included, please clarify, and modify Fig. 1 accordingly to reflect this information.

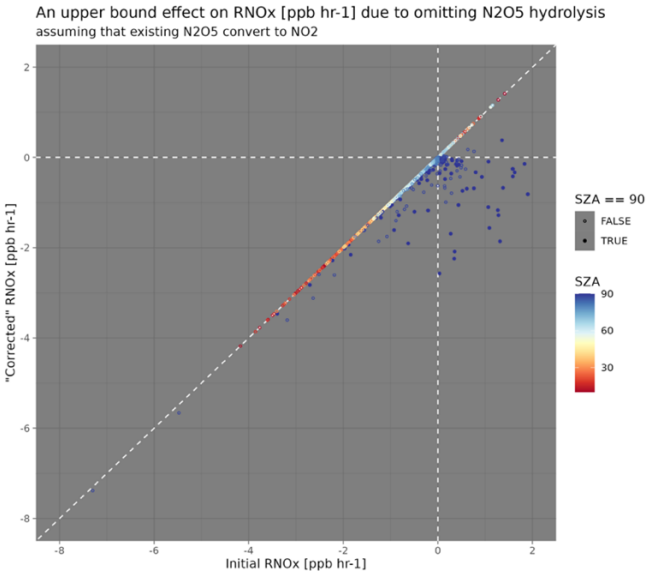


The gas-aerosol chemistry of NOx was not turned on in WRF-Chem as properly addressing such reactions requires knowledge about aerosol loading and composition as well as uncertainties in the dependence of reaction probability on water vapor and temperature (e.g., Real et al., 2008), which remain challenging topics. Certainly, the N2O5 hydrolysis is an important NOx pathway during nighttime. Essentially, we did not enable the arrow from N2O5 to HNO3 given our choice of WRF-Chem (**see simplified diagram above**). Note that RACM considers the thermal decomposition of N2O5 (Stockwell et al., 1997).

As a result, more NOx will survive the night and appear in the morning due to the photolysis and thermal decomposition of N2O5 instead of being partially lost to HNO3. The omission of N2O5 hydrolysis generally causes a high bias in NO_2 concentration and a resulting slow bias in the chemical loss of NOx over urban environments. In other words, the NO_x chemical tendency (RNOx = PNOx- LNOx) is larger in the atmosphere (i.e., faster NOx loss rate) than in the model.

To address this comment, we added a simple sensitivity experiment to understand the impact of the omission of N2O5 hydrolysis via simple sensitivity analysis. For this experiment, we simply

assume that all existing N2O5 in the current setup of WRF-chem photolyzed into NO_2 during the day (SZA < 90), despite those N2O5 being extracted from the non-hydrolysis runs. We then update the bin-averaged NOx chemical tendency and compare it with the original. Reduction in chemical tendency is found for each 2-deg SZA bin and a maximum reduction occurs at SZA of ~90 degrees (**blue dots in the figure to the right**).



Nonetheless, considering the local overpass time of TROPOMI of ~1 pm, we expect the impact to become progressively small as the plume disperses. In other words, it becomes more of a background issue/bias, which may not greatly affect the future emission estimates if the background is properly subtracted from both TROPOMI and the model.

2. The NOx chemical tendency is parameterized as a function of NOx concentration and the solar zenith angle. It is unclear why these were the only two variables chosen and whether they account for most of the variation in the NOx chemical tendency. Knowing the fraction of variation explained by these variables would be useful. I expected temperature (as a proxy for seasons) and the NO2/NO ratio to be potentially important variables.

We agree that we did not explicitly show to what degree the variability in NOx chemical tendency is explained by the chosen two variables. Although adding feature variables to the NOx curves will necessarily explain more variability of NOx chemical tendencies (RNOx) when training RNOx, some feature variables are difficult to "obtain" or only have a marginal effect in predicting RNOx. Our rationale is to choose environmental variables compared to atmospheric concentrations of other tracers like ozone unless possible "proxies" are available for those tracers' abundance.

To justify our choices and inform the "fraction of variation explained by these variables" as suggested by the reviewer, we conducted sensitivity analyses. The grid-level RNOx is grouped and averaged into bins of feature variables so that the variability in initial RNOx will be damped after grouping. Here we compared the initial RNOx derived from WRF-chem and the bin-averaged RNOx and reported their Pearson correlation coefficient and RMSE (**see figure shown below, now as Fig. S2a and S2b in the revised manuscript**). Those two statistical metrics were reported for every choice of variables and four scenarios with higher vs. lower NOx levels (orange vs. blue bars) and day vs. nighttime (empty bars vs. bars with stripes).

The variability in RNOx is better preserved as more variables are considered especially when NOx is high, i.e., >= 1 ppb (**orange bars in Fig. S2a**). Although the correlation remains low for lower NOx conditions (< 1 ppb), the chemical tendencies stay low, leading to small absolute random errors (**Fig. S2b**). Nonetheless, including NO_2/NO ratio and VOCR seems to have marginal impacts

on preserving RNOx variability over polluted regions. Introducing ozone concentration would help, but the prediction of ozone is arguably more complicated than the prediction of NOx. The correlation coefficient improves when adding air temperature, but RMSE does not largely reduce.

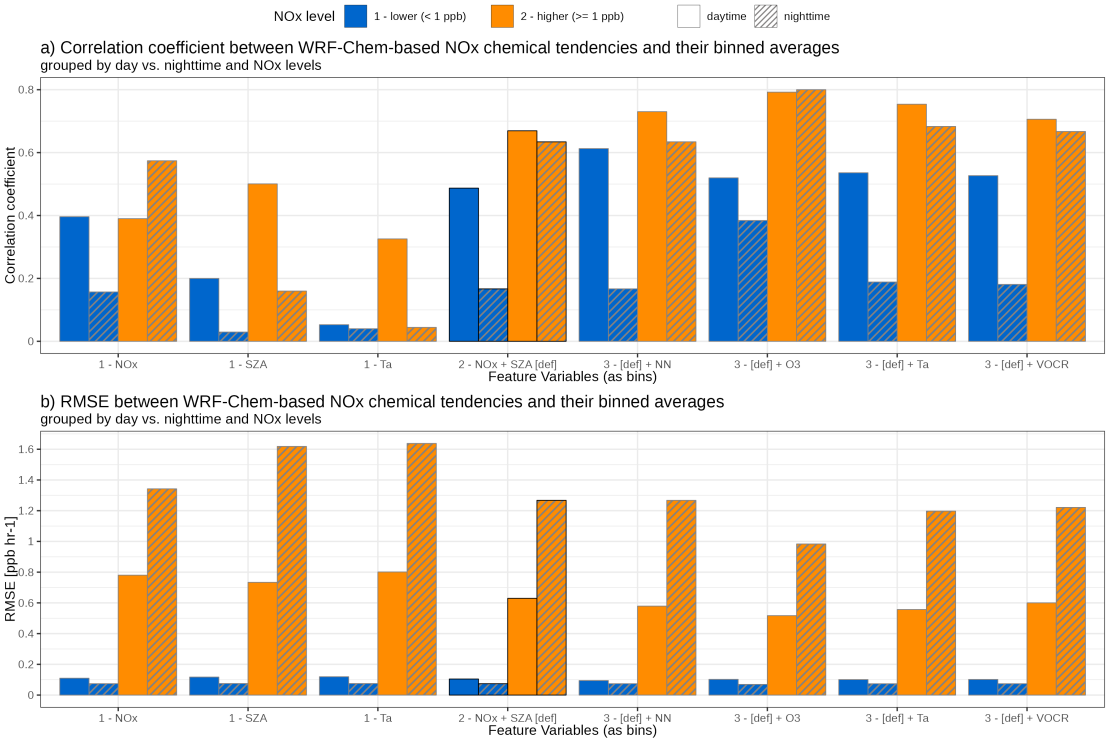


Figure S2 (ab) Correlation coefficient and RMSE between the raw RNOx directly derived from WRF-Chem and the bin-averaged RNOx based on 8 combinations of feature variables. The number in front of the feature variables on the x-axis denotes the total # of variables used when grouping the raw RNOx. The combination we used (2-NOx+SZA for NOx concentration and solar zenith angle, SZA) is highlighted in bars with black outlines. Additional variables tested include air temperature (Ta), NO_2-to-NOx ratio (NN), ozone (O3), and VOC reactivity (VOCR). Results are reported separately for higher or lower NOx levels (orange or blue) during the day or night (empty bars or bars with strips).

In conclusion, we chose two variables in this version: 1) NOx concentration for initiating its non-linear dependence on chemical tendency, and 2) SZA for indicating photolysis rates. We have acknowledged that these NOx curves/relationships can be improved and replaced using more variables or different chemical schemes. In particular, the addition of background ozone if properly constrained would be an important advance (which also helps constrain the NO2-to-NOx ratio). **We now added a brief explanation in Sect. 2.1:**

The above grouping procedure of R_{NO_x} based on a finite number of bins of C_{NO_x} and θ unavoidably reduces the variability of R_{NO_x} that were directly derived from WRF-Chem. To assess the extent to which the R_{NO_x} variability can be explained by the selected binning feature variables, we performed a sensitivity test to quantify the deviation of bin-averaged R_{NO_x} from the initial R_{NO_x}. Generally, the R_{NO_x} variability is better preserved over polluted regimes with higher NO_x level > 1 ppb than over low-NO_x regimes (**Supplement Figs. S2ab**). Choosing C_{NO_x} alone better explains the R_{NO_x} variability than choosing SZA or air temperature alone. Including additional variables (e.g., air temperature, NO_2–to–NO_x ratio, and VOC_R) on top of our default choice of SZA and C_{NO_x} marginally improves the prediction of R_{NO_x} except for the inclusion of ozone. However, estimating ozone remains a challenging problem, thereby ozone is not included as a feature variable in this study.

3. The calculation of the NOx chemical tendency was based on WRF-Chem simulations for three cities: Los Angeles, Shanghai, and Madrid. The rationale behind selecting these specific cities seems arbitrary. They are unrelated to the power plants and cities that were chosen for model evaluation. It would be helpful to have an explanation for this choice.

LA, Shanghai, and Madrid represent typical megacities in North America, Asia, and Europe and are chosen given their distinct sectoral emissions of GHGs, NOx, and VOCs. The fact that the three training cities don't overlap with the targets for predictions is in turn a benefit, which avoids possible overfitting of the NOx tendency relationships with feature variables. If we trained and then tested over the same three cities, it would be less convincing in terms of the generalization of the model parameterizations. We add a brief explanation In Sect. 2.1:

~~Considering urban areas being the main focus of our study~~Focusing primarily on polluted environments, we carried out WRF-Chem simulations ~~over~~ for three mid-latitude cities and ~~focused on outputs over~~ extracted model outputs from a 2° ×2° region ~~around the citycenter to spotlight chemical regimes in urban environments. The three citiesare~~ centered around each city. Three cities, namely Los Angeles in the US, Shanghai in China, and Madrid in Spain ~~, which varied in and VOC emissions and climatology.~~ represent typical megacities in North America, Asia, and Europe. Their varied climatic conditions and sectoral emissions of NO_x, VOC, and GHGs provide a holistic view of the variability of NO_x chemical tendency. While our analyses extended to power plants and cities beyond these three training sites when compared to TROPOMI data (**Sect. 4**), it helps assess the broader applicability of our chemical parameterizations.

4. The assumption of the NOx chemical tendency being independent of height (Eq. 2) is not accurate considering the vertical gradients of NOx near the surface during nighttime and early mornings. It seems important to discuss any limitations arising from this.

We apologize for the misleading text and Eq.2 and clarify that NOx chemical tendency (RNOx) is not assumed to be independent of height. Since RNOx relies on NOx concentration (CNOx) and CNOx varies vertically, RNOx varies with altitude. We clarify that both RNOx and CNOx were extracted from each WRF-Chem model grid and from 12 levels within the boundary layer. Those RNOx and CNOx were further grouped up to create those NOx curves in Figure 3. When it came to predicting CNOx in STILT-NOx, modeled NOx concentrations vary between STILT air parcels at different vertical levels because STILT footprint takes care of the atmospheric transport, and RNOx is further prescribed as a function of NOx concentration. We have now modified Eq. 2 and the relevant text:

By leveraging WRF-Chem's chemical diagnostic ~~feature~~capability, we derive the net chemical tendency of NO_x within each hour [R_{NO_x}, ppb hr^{-1}] ~~per model grid~~ for every model grid within the lower 12 vertical levels (x, y~~) based on the output cumulative chemical~~ , z). R_{NO_x} is calculated specifically from the cumulative changes in NO and NO_2 concentrations solely due to chemical reactions (i.e., "chem_no2" and "chem_no" in WRF-Chem registry) following **Eqs. 2**:

$$\sum_{h_0}^{h} \Delta C_{NO_x}(x,y,z) = \sum_{h_0}^{h} \Delta C_{NO}(x,y,z) + \sum_{h_0}^{h} \Delta C_{NO_2}(x,y,z) \quad (2a)$$

$$R_{NO_x}(x,y,z,h) = P_{NO_x}(x,y,z,h) - L_{NO_x}(x,y,z,h) = \frac{\sum_{h_0}^{h} \Delta C_{NO_x}(x,y,z) - \sum_{h_0}^{h-1} \Delta C_{NO_x}(x,y,z)}{1\ hr} \quad (2b)$$

where model hour h denotes the ~~beginning time of the hour interval of the hourly WRF-Chem outputs .~~ start time of each hour interval in the WRF-Chem outputs and z denotes the index of model vertical levels (i.e., from 1 to 12). $\sum_{h_0}^{h} \Delta C_{NO_x}$ describes the cumulative net changes to NO_x concentration given chemical reactions from the initial model hour h_0.

5. It would be useful to assess the consistency of the NOx chemical lifetime from WRF-Chem to the available observations, although limited.

We agree that assessing WRF-Chem against observations would be useful. However, the main goal of this study is not to determine if the RACM2 chemical scheme in WRF-Chem is correct, but to investigate if a simplified model can replace a sophisticated chemical scheme, assuming it is correct. There had been numerous studies aiming to evaluate and improve WRF-Chem chemical schemes. Lastly, we recognize that our simplified NOx curves can be improved and replaced.

NO2-to-NOx ratio (Sect. 2.2):

1. The reactions of NO + HO2 and NO + RO2 to form NO2 are excluded, but they are important in the boundary layer.

We agree that RO2/HO2 are additional sources to oxidize NO, which is missing from the estimate of the NO_2-to-NOx ratio (that relies on modeled NOx values and a prescribed Ox level of 50 ppb, which neglects local Ox variability with VOC reactivity as discussed in Sect. 5.3.1). Properly resolving such reactions requires knowledge/input of VOC emission and chemistry, which would be even more challenging and essentially requires a full chemistry model. Yet, there may be room for simple parameterizations as further discussed in Sect. 5.3.1, including 1) adding background ozone, tracking ozone or its production rate along trajectories by leveraging satellite column observations of HCHO, and 2) parameterizing the NO_2-to-NOx ratio as a function of feature variables including VOCR. Relevant modifications to Sect. 5.3.1 are attached as follows:

VOC_R may also affect the O_x level ~~, thus, the~~ and the NO_2–to–NO_x ratio. The prescribed O_x level of 50 ppb ~~neglected the~~ (**Sect. 2.2**) overlooks the nonlinear O_x variability ~~given its similar non-linear dependence on concentrations and~~ related to VOC_R (Murphy et al., 2007; Li et al., 2022). ~~Specifically,~~ In NO_x-limited scenarios, OH favors the oxidation of VOCs, and local-scale O_x is ~~primarily~~ predominately produced by NO + RO_2 or ~~NO +~~ HO_2 ~~under limited-rich conditions. Consequently, higher~~ , suggesting higher O_x levels with increased NO_x concentrations~~result in higher levels. However~~. The omission of NO + RO_2 or HO_2 in **Eqs. 3** could lead to an underestimation of the NO_2–to–NO_x ratio, which likely explains the modeled tNO_2 being consistently lower than observations over background regions. Conversely, under NO_x-saturated conditions, the ~~production of~~ consumption of OH by NO_x may limit the VOC oxidation and O_x ~~is suppressed by~~production, leading to a decline in O_x level as NO_x concentration ~~increases. For example,~~ rises. Consequently, the NO_2–to–NO_x ratio might be overestimated when true O_x levels ~~may be lower than~~ fall below 50 ppb ~~at saturated regimes. If a simplified~~ (particularly under stagnant atmospheric mixing) or underestimated due to the absence of NO + RO_2 reactions. Nevertheless, our pre-determined O_x ~~-nonlinearity is implemented, the conversion from NO to would be further slowed to alleviate strong overestimation of tropospheric under stagnant mixing conditions. Nonetheless, our prescribed~~ level of 50 ppb ~~serves~~ acts as a first-order ~~cap to prevent the endless non-physical conversion~~ limit to prevent unrealistic conversion from NO to ~~from NO when~~ NO_2 at extremely high NO_x ~~is extremely high since~~ levels when O_3 ~~can be~~ is being titrated.

To address these limitations, one potential approach is to leverage formaldehyde concentrations retrieved from TROPOMI. Recent studies revisited the use of the formaldehyde-to-NO_2 ratios (i.e., FNR) from satellites as a means of inferring O_3 production rates (Goldberg et al., 2022; Souri et al., 2022). Our WRF-Chem simulations, which were used to parameterize the NO_x chemical tendency, show that modeled formaldehyde generally increases with VOC_R with varying slopes influenced by SZA and NO_x concentrations (**Supplement Fig. S18c**); and O_3 concentrations scale non-linearly with FNRs with O_3 concentration approaching a background value at high FNR > 10 (**Supplement Fig. S18d**). Even though satellite-based FNRs may theoretically help probe O_3 or O_x concentration to better parameterize NO_2–to–NO_x ratios, Souri et al. (2022) stressed that retrieval errors especially from formaldehyde (40 to 90% with $\leq$ 50% over cities) and inherent chemical errors of the predictive power of FNRs may hinder the broad application of space-based FNRs at the current stage. Nonetheless, sensitivity analyses in **Sect. 3** indicate an overall chemical uncertainty in tNO_2 of about 10 to 20% with respect to NO_2 signals, even if perturbed O_x level is much lower than 50 ppb (**Fig. 4**).

2. Line 242: Please clarify how the NOx chemical tendency in the model change when ozone is titrated near high emitters.

We clarify that in STILT-NOx, only NOx (no ozone) is tracked and updated per timestamp along the trajectory; and the calculation of NOx chemical tendency relies on NOx concentrations and SZA. Thus, RNOx would not change as ozone decreases in the model. See our sensitive analysis above about the performance of estimating RNOx using additional variables (see our earlier response to the second comment).

The initial text was to explain how the NO_2-to-NOx ratio parameterized separately from RNOx would decrease when NOx emission/concentrations become extremely high (titrating ozone). Such titrations near high NOx emitters are controlled by the prescribed total oxidant level, [Ox] = [O3] + [NO_2], despite a simple constant value being assigned (see response above for limitation of this constant Ox level). For example, if a STILT-NO_x air parcel passes by a larger emitter like a power plant, its tagged NOx concentration will be largely enhanced but its final NO_2 column at the receptor/sounding location may stay low given limited oxidant capacity.

Eq 1: The processes of NO2 dry deposition and mixing between the mixed layer and the free troposphere seem to be neglected.

We thank the reviewer for pointing out those physical processes. We agree that neglecting the dry deposition of NO_2 or NO may overestimate their concentrations and alter the estimated chemical tendency of NOx along the trajectories that further feedback to the concentrations. Previous studies implemented the deposition module that relies on the key estimation of "dry deposition velocity" for air parcels residing within the lower 50 m from the surface layer (e.g., Wen et al., 2011). Moreover, the mixing between the mixed layer and the free troposphere (FT) was neglected and requires future model modifications, such as 1) quantification of the entrainment zone and 2) the choice of vertical mixing time-/length- scales at the PBL-FT interface. Although issues with mixing and deposition were not fully resolved in this study, we acknowledge these limitations and discuss possible future improvements in Sect. 5.3.2:

5.3.2 ~~The impact from emission profiles and inter-parcel mixing scales~~Uncertainties in non-chemical processes

Besides simplification of chemical reactions, modeled tNO_2 values can be subject to a few physical processes and parameters, including emission profiles, inter-parcel mixing scales, and dry deposition.

The underlying STILTv2 (Fasoli et al., 2018) accounted for a gradual growth of the mixed layer height over the hyper-near-field area around emissions. ~~Convolving~~ Yet, by convolving the STILT footprint with NO_x emissions, we assumed that emissions ~~originating~~ originate from the surface and are uniformly mixed over the mixed layer without considering the possible uneven distribution of emissions from different vertical levels. In reality, under stable atmospheric conditions, the stack heights or plume heights of emission sources can sometimes extend above the shallow PBL. Our current assumption may thus lead to an overestimation in modeled concentrations, and such biases can in turn affect the estimate of NO_x tendency. More importantly, changes in the vertical ~~emission profile~~ profile of emissions can lead to changes in concentration per model level, which ~~also~~ affect the tropospheric ~~column results~~ columns as the typical averaging kernel profile is far from uniform within the PBL. Recall that TROPOMI NO_2 AKs decreases rapidly towards the surface (**Fig. 1**). Hence, placing ~~a~~ an emission plume at the surface or ~~at~~ an elevated altitude (e.g., 400 m) can cause a discrepancy in ~~the modeled column values.~~ modeled column concentrations. In

addition, if the wind shear is strong over an intensive point source (likely the Intermountain example in **Fig. 5c**), assumptions in the injection height and vertical profile of emission plumes may affect the modeled plume shape and possibly deviate the estimated near-field wind bias following **Sect. 5.2.1**. Noticeably, Maier et al. (2022) ~~implemented~~ investigated the influence of inaccurate representation of emission profiles on the flask-like modeled concentrations by implementing a time-varying sector-specific emission profile into STILT. ~~Yet, the influence~~ Such an impact on column concentrations ~~due to changes in emission profiles may require more~~ may be minimized but yet requires future in-depth investigations, particularly over point sources.

Accounting for inter-parcel mixing was an important aspect when developing Lagrangian chemical models. Omitting inter-parcel mixing makes solving for non-linear processes (such as chemical NO_x loss) problematic. On the contrary, Eulerian models suffer from excessive numerical diffusion. Mixing that is too strong smooths the spatial gradient of concentration and can lower the concentration within the fixed model grids, which may cause slight shifts in NO_x regimes. Valin et al. (2011) suggested that a spatial resolution of 4 to 12 km is sufficient to capture the non-linearity in NO_x loss rate. As for Lagrangian models, efforts can be made to enable the flux exchange between air parcels via deformations (Konopka et al., 2019; McKenna et al., 2002). ~~STILT realized the~~ In addition to the inter-pacel mixing within the mixed layer (ML), several other turbulent mixing processes require future investigation, including (1) horizontal mixing in the free troposphere (FT), (2) vertical mixing between the ML and FT, and (3) mixing between tracked air parcels with the untracked surrounding background. For example, Real et al. (2008) utilized a linear relaxation with exponential decay of the plume concentrations towards the background based on a timescale of 2 days to address the third mixing process. The second mixing process requires future modifications involving the determination of entrainment zones and mixing hyperparameters for such ML-FT exchange.

The original STILT model realized vertical mixing by diluting surface emissions ~~over~~ across the ML height (Lin et al., 2003) ~~. We~~ and we further enabled an exchange in pollutants' concentrations with prescribed mixing length- and ~~time- scales following (Wen et al., 2012). As a final sensitivity test~~time-scales representing typical horizontal mixing rates (**Sect. 2.3**). As final sensitivity tests, we simulated tNO_2 ~~using~~ based on a spectrum of mixing hyperparameters ~~over~~ for the New Madrid power plant~~and found minimal influence on modeled values per sounding (uncertainty < 20%, .~~ Uncertainties in the mixing parameters result in minimal uncertainties on the sounding-level modeled tNO_2 values (**Supplement Fig. S19**). For example, differences in modeled tNO_2 between the mixing and non-mixing simulations become larger as mixing becomes faster and for receptors/soundings located on the edge of the plume (i.e., only a small fraction of the trajectories encountered power plant emission in **Supplement Fig. S20**). Uncertainties in the prescribed mixing hyperparameters contribute even less to the modeled values over urban areas (i.e., < 10% for Phoenix cases), where emissions are generally better mixed than at power plants. In addition, such mixing influence can vary with the spatial resolution of the emission inventory used in the simulations.

The dry deposition of NO_2 was not factored into this study, which could lead to an overestimation of modeled NO_2. For future model implementations, it is possible to track loss of NO_2 concentrations due to dry deposition by calculating "dry deposition velocities" (e.g., **?**) when air parcels descend close to the surface, e.g., 50 meters above the surface (Wen et al., 2012).

Eq. 5 assumes that the model transport and chemistry errors are independent, when in fact these errors are related.

While it is difficult to fully separate the transport error from the chemistry error (since the physical and chemical processes are fully coupled), our transport error analysis that involves perturbing wind fields and recalculating [NOx] along trajectories implicitly contains errors in tNO2 due to errors in wind and chemistry. This can be explained as follows:

Imagine a modeled air parcel/trajectory passing over a point source in the default run. With a small perturbation to the model winds, the air parcel no longer passes over that point source. The NOx concentration of these two air parcels would be different for two reasons:

1) the wind fields are perturbed → leading to changes in emissions and NOx concentrations (e.g., lower values for the perturbation run); and
2) the associated feedback involving chemistry → lower NOx concentration in the end simulation leads to changes in chemical tendency (and so on for further timesteps).

Our linear combination of three error sources would likely have been a conservative estimate given the nature of the transport error analysis. An alternative approach is to perturb the wind fields and the chemistry simultaneously to account for the covariance between the two sources of uncertainties, which is hard to realize with existing tools.

Fig. 5 and elsewhere: Please clarify how tNO2, the tropospheric NO2 column, is converted to a volume mixing ratio.

Thanks for this comment. We clarify that "tNO2" denotes the tropospheric NO_2 mixing ratio that is derived from the initial tropospheric NO_2 vertical column density (VCD). **We now add a brief explanation to the introduction section when tNO2 is first brought up:**

In this study, we present a non-linear modeling framework, STILT-NO_x, to simulate tropospheric column-average NO_2 ~~columns~~ mixing ratio (tNO_2) as retrieved from TROPOMI. Note that initial NO_2 vertical column density [VCD, molec cm^{-2}] is converted to tNO_2 [ppb] by dividing by a dry air VCD. The dry air VCD is calculated by integrating a profile of the ideal gas number density of air minus a modeled water vapor profile.

Line 382: Considering the model's sensitivity to errors in the wind direction and speed, it may be better to use winds from reanalyses datasets or to bias correct the model using nearby observations. This seems important for the inversion work planned in the future. Are there other ways to reduce this error?

Yes – as emphasized in Sect. 5, wind biases affect the interpretation of the model-data alignment of column NO_2 and other species and possibly the corresponding inverted fluxes/emissions. We note that meteorological reanalyses with relatively coarse spatiotemporal resolutions are not real observations and are associated with uncertainties. For example, the GFS meteorological field driving STILT is a data assimilation (reanalysis) product but still cannot resolve highly fine-scale wind patterns. Moreover, the nearby observations are spatially limited, e.g., the radiosonde observations used for regional wind assessment and transport error estimates (Sect. 3).

We may further differentiate random wind errors from systematic wind biases. The former error can be quantified (Sect. 3) and usually be propagated into the observational error matrix during the inversion. It is the wind biases that have hardly been addressed in inversion studies. Possible approaches as we discussed in Sect. 5.2.1 include the rotation of plumes or a more sophisticated data assimilation approach to correct emissions and wind altogether (e.g., Liu et al., 2017). In the ongoing inversion work, we will design and conduct OSSE to investigate the quantitative impact of wind biases on inverted fluxes and explore ways to explicitly address the wind error.

Lines 564-570: These lines are unclear.

As shown in Fig. 1, the two essential competing loss pathways of NOx are NO_2 + HO-> HNO3 and NO + RO2-> RONO2 with a minor branching ratio (typical value of 0.04 and further depends on

the R groups). In other words, when VOCR stays high and NOx is limited, there is a tendency to form RONO2 over HNO3, which explains why the trough of the NOx curves shifts to the higher end of NOx concentration as shown in Fig. 10. We have reworded these lines as follows:

When considering lower SZAs (consistent with TROPOMI overpass time of 1 pm ~~TROPOMI observations~~local time), the general non-linear ~~shape~~ characteristic of these NO_x curves holds as VOC_R increases (**Fig. 10**). Higher VOC_R relative to ~~the~~ lower NO_x concentration favors the oxidation of VOCs ~~and the~~ by OH and the associated minor loss pathway of NO + RO_2 to form alkyl nitrates with a ~~small branching ratioover the~~ minor branching ratio, over the competing major NO_x loss pathway of NO_2 + OH (**Fig. 1**). ~~To compete with the VOC oxidation in reacting with OH~~With rising VOC_R, the NO_x ~~loss tendency becomes slower (~~chemical tendency becomes more positive (P - L, **Supplement Fig. S18a**) and ~~concentrations at the optimal point (where two loss pathways reach their maximums) must be increased, as illustrated by the increasing from 1.4 to 3.7 ppb~~ the net loss timescale elongates (e.g., ts_min from 2 to 4 hours in **Fig. 10**~~. For the same reason, the net loss timescale (ts_{NO_x}) generally rises as increases,~~). Moreover, NO_x is required to reach a higher level to compete with the reactions involving VOCs, evident by the shift in the trough of the NO_x curves (e.g., ~~from about 2 to 4 hours~~NOx_min from 1.4 to 3.7 ppb in **Fig. 10**). To put it in context, the NO_x curves shown in **Fig. 3** represent ~~the typically curves under moderate~~ typical patterns as long as VOC_R ~~(e.g., <~~ remains below 10 s^{-1}~~). Elevations in ts_{NO_x} with may be problematic when NO is high because the ts_{NO_x} has already been quite high; while more erroneous under conditions with moderate NO and extremely high of over 10 s^{-1}~~.

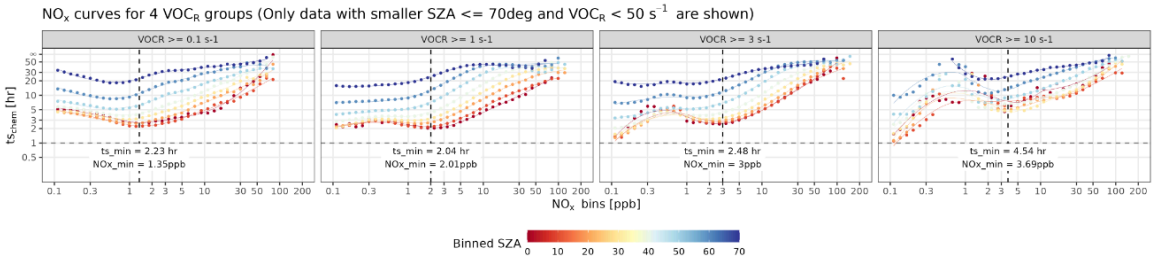


Figure 10. (a) Similar to **Fig. 3b**, but differentiated by 4 intervals of VOC_R and SZA bins smaller than 70 degrees with ~~an~~ a spacing of 10 degrees. All panels here utilized model results from the same WRF-Chem simulations described in Sect. 2 and **Appendix A**.