Title
Uncertainties in climate assessment for the case of aviation NO

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Nitrogen oxides emitted from aircraft engines alter the chemistry of the atmosphere, perturbing the greenhouse gases methane (CH₄) and ozone (O₃). We quantify uncertainties in radiative forcing (RF) due to short-lived increases in O₃, long-lived decreases in CH₄ and O₂, and their net effect, using the ensemble of published models and a factor decomposition of each forcing. The decomposition captures major features of the ensemble, and also shows which processes drive the total uncertainty in several climate metrics. Aviation-specific factors drive most of the uncertainty for the short-lived O₃ and long-lived CH₄ RFs, but a nonaviation factor dominates for long-lived O₂. The model ensemble shows strong anticorrelation between the short-lived and long-lived RF perturbations (R² = 0.87). Uncertainty in the net RF is highly sensitive to this correlation. We reproduce the correlation and ensemble spread in one model, showing that processes controlling the background tropospheric abundance of nitrogen oxides are likely responsible for the modeling uncertainty in climate impacts from aviation.

Quantifying uncertainties in climate processes is a major priority for climate research, as exemplified by the widespread use of probability distributions and uncertainties in the Intergovernmental Panel on Climate Change (IPCC) Fourth Assessment Report (1). To understand and attribute changes in the greenhouse gases and aerosols that force climate change, we rely on chemistry-transport models (CTMs), which require further characterization of their errors (e.g., 2–5). Aviation emissions are estimated to contribute 5% of anthropogenic climate forcing, with a nearly threefold uncertainty (6). This forcing occurs mainly through aviation-induced cloudiness and emissions of CO₂ and nitrogen oxides. Climate impacts of other aviation emissions—CO, SO₂, soot, hydrocarbons, and water vapor—are highly uncertain, but estimated to be much smaller (6). In this paper we quantify one type of climate forcing and its uncertainty, the impact of aviation emissions on the greenhouse gases ozone (O₃) and methane (CH₄).

Aircraft engines emit NO and NOₓ (NOₓ), which are indirect greenhouse gases through their chemical reactions impacting O₂ production and the CH₄ lifetime. As an O₃ precursor, NOₓ emissions increase the tropospheric column of O₃ and its radiative forcing. Aviation NOₓ emissions have a stronger impact on O₃ than surface emissions on a per molecule basis because they occur at high altitudes (7). Previous model-based estimates of the amount of O₃ generated by aviation NOₓ emissions vary by up to 100% and these differences have been attributed to a range of causes: e.g., the ratios of NO: NOₓ, and OH:HO₂, background NOₓ levels (8), location and time of emissions (9–11), the amount of sunlight (12), and atmospheric mixing (13, 14).

Early work on climate forcing by aviation assumed that the impacts would be short-lived and limited to the northern latitudes containing the major flight corridors because of the short lifetime (months) of O₃. Subsequent studies, however, found that the short-lived, aviation-induced O₃ and CO perturbations propagate to the tropics where they boost OH levels and trigger a long-lived negative perturbation to global CH₄ abundance (15). An equally long-lived, global, and negative O₃ perturbation accompanies the CH₄ perturbation, and is based on numerous studies showing that the increase in mean tropospheric O₃ since preindustrial times is partially due to the increase in CH₄ (16, 17). The negative aviation radiative forcing (RF) due to long-lived CH₄ and O₃ perturbations nearly cancels the short-lived positive RF due to O₃ (12), which creates large uncertainty in the net forcing. Although most studies find that aviation NOₓ emissions cause net positive RF, some find negative forcing (8, 11, 18). This work aims to quantify the RF uncertainty using two complementary approaches. Part of the analysis is similar to that in earlier studies and IPCC assessments: We look at the ensemble of published model results and perform sensitivity analyses on poorly understood processes with a single CTM. We also develop another analysis here: We decompose the RF into components and factors, assess the possible errors in each, and then propagate the final uncertainty in RF. Taken together, these approaches point to atmospheric measurements that would reduce uncertainties in key processes involved in climate forcing.

**RF Uncertainty from Model Ensembles**

In this section we analyze aviation NOₓ RF from the ensemble of all published studies since the IPCC aviation assessment (15), which synthesized earlier work. Fig. 1 shows each RF estimate from the last 12 y, broken down into its short-lived and long-lived O₃ and CH₄ components. These estimates come from diverse models, including CTMs (7, 10–12, 18–21) and coupled chemistry-climate general circulation models (GCMs) (7, 15, 19–21). They are based on a wide range of scenarios for aviation and other emissions. Studies find little evidence for nonlinearity of the atmospheric response to aviation emissions over the range 0.4 to 1.3 Tg(N) a⁻¹ used in the ensemble here (7, 18), so we normalize all results to 1 Tg(N) a⁻¹ from aviation.

The RF values in Fig. 1 and throughout this work are the steady-state responses to aviation NOₓ emissions. Most studies calculate these by the difference between two steady-state simulations with differing aviation NOₓ emissions (7, 18–21). Three studies report 100-year integrated RF responses to pulse emissions of aviation NOₓ (10–12), which are comparable to the steady-state response because the integration time is much longer than the CH₄ perturbation lifetime (10–15 y). The latter studies use pulse durations between 1 mo and 1 y, which could conceivably increase the spread of RF estimates, but we find their results to be within the envelope of RF from steady-state simulations. Our ensemble statistics would not change by excluding them. We exclude one recent aviation study (22) because its results derive from the same model simulations as another included study (21) and another because of its different experimental design (23). Stated RFs are based on composition changes in the troposphere and do not include stratospheric response, although some studies also include the lower stratosphere, where aviation NOₓ also increases O₃.

Author contributions: C.D.H. and M.J.P. designed research; C.D.H. and Q.T. performed research; C.D.H., Q.T., and M.J.P. analyzed data; and C.D.H. and M.J.P. wrote the paper.

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Past efforts to quantify the uncertainty in the climate impacts of aviation NOx have relied primarily on ensembles of model simulations, as in the prior section. In this section we present an alternative method for determining the uncertainty of aviation impacts based on decomposing the RF into its key factors. We use literature review and expert judgment to assess the uncertainty in each factor and propagate these to the uncertainty in RF. Many of the factors can be derived from the general climate forcing literature and do not require model simulations that are specific to aviation. A major advantage of this approach is that we can identify the dominant sources of uncertainty.

The steady-state, global mean RF from aviation NOx emissions ($F$) can be broken into components due to the short-lived O$_3$ response ($F_{\text{short O}_3}$), and the long-lived CH$_4$ and O$_3$ responses ($F_{\text{long CH}_4}$, and $F_{\text{long O}_3}$, respectively):

$$ F = F_{\text{short O}_3} + F_{\text{long CH}_4} + F_{\text{long O}_3}. $$

These components can in turn be factorized into key terms summarizing the chemical interactions between NOx, O$_3$, and CH$_4$, including their feedbacks:

$$ F_{\text{short O}_3} = (dO_3/dE) (dF/dO_3), $$

$$ F_{\text{long CH}_4} = (dL_{\text{CH}_4}/dE) (f_{\text{CH}_4}[dF/d\text{CH}_4]), $$

$$ F_{\text{long O}_3} = (dL_{\text{O}_3}/dE) (f_{\text{O}_3}[dF/d\text{O}_3]), $$

where $dO_3/dE$ is the short-lived O$_3$ response to aviation NOx emissions ($E$) and $dL_{\text{CH}_4}/dE$ is the accompanying relative change in CH$_4$ lifetime, $dF/dO_3$ and $dF/d\text{CH}_4$ are the RF efficiencies of tropospheric O$_3$ and CH$_4$, $[\text{CH}_4] = 1.78$ ppm is the global mean CH$_4$ abundance, $f_{\text{CH}_4}$ is the feedback factor, which prolongs the lifetime of CH$_4$ perturbations (16), and $dO_3/[d\text{CH}_4]$ is the O$_3$ component of the long-lived CH$_4$ perturbation (26). Note that the product of the first three terms on the right-hand sides of Eqs. 3 and 4 equals the steady-state response of CH$_4$ abundance to aviation emissions (14).

Table 1 reviews the best estimates and one-sigma uncertainties (68% confidence interval) of each factor in Eqs. 2–4, including the values adopted in this work. Atmospheric measurements provide no direct constraints on these factors. Although the range of models does not encompass all potential uncertainties and errors, it does in this case provide a large ensemble.

Only two of the factors—$dO_3/dE$ and $dL_{\text{CH}_4}/dE$—require model simulations specific to aviation to derive their values and uncertainties, although $dF/dO_3$ is also somewhat dependent on the aviation perturbation (see below). The values adopted here for $dO_3/dE$ and $dL_{\text{CH}_4}/dE$ encompass recent estimates based on model ensembles (19, 20), but exclude values from IPCC (15) that were based on early CTMs, which had coarse resolution at aircraft altitudes, and were not supported by subsequent studies. The poorly understood sources of model differences in these terms are discussed below, but likely involve the range of NOx: NO$_2$ and OH:HO$_2$ ratios and background NOx abundances across the models.

The remaining factors can be estimated from general literature on atmospheric composition and climate, without requiring model studies or calculations specific to aviation. The O$_3$ response to long-lived CH$_4$ changes, here $dO_3/[d\text{CH}_4] = 3.5 \pm 1.0 \ DU/(\text{ppm CH}_4)^{-1}$, can be calculated from models (17, this work) and is qualitatively supported by numerous studies showing that the increase in mean tropospheric O$_3$ since the preindustrial era is partially due to the increase in CH$_4$ (16). Our estimated range for the CH$_4$ feedback factor ($f_{\text{CH}_4} = 1.4 \pm 0.1$) includes values from models with widely varying chemical mechanisms (16, 17, this work). The final terms involve mapping the changes in green-
from studies of the preindustrial to present-day changes in O$_3$ derived from aviation-specific model tests (7) overlaps that
between the short-lived and long-lived O$_3$ significantly between models. The RF efficiency may also differ
well-known because the perturbation patterns themselves vary
value here,

DU and

Table 1. Factors governing the climate impact of aviation NOx emissions

<table>
<thead>
<tr>
<th>Factor, unit</th>
<th>Value</th>
<th>Uncertainty*</th>
<th>Source*</th>
</tr>
</thead>
<tbody>
<tr>
<td>$d[O_3]_{short}$/$dE$</td>
<td>0.9 ± 25%</td>
<td>(15)</td>
<td></td>
</tr>
<tr>
<td>DU($O_3$)</td>
<td>0.6 ± 15%</td>
<td>(19)</td>
<td></td>
</tr>
<tr>
<td>$d[O_3]_{long}$/$dE$</td>
<td>0.51 ± 36%</td>
<td>(20)</td>
<td></td>
</tr>
<tr>
<td>$d[CH_4]$/dE</td>
<td>0.74 ± 0.15</td>
<td>Adopted</td>
<td></td>
</tr>
<tr>
<td>%($O_3$/Tg aviation N) a$^{-1}$</td>
<td>0.60 ± 20%</td>
<td>(15)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>-1.8 ± 36%</td>
<td>(19)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>-1.55 ± 38%</td>
<td>(20)</td>
<td></td>
</tr>
<tr>
<td></td>
<td>-1.7 UCI</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>-1.7 ± 0.35</td>
<td>Adopted</td>
<td></td>
</tr>
<tr>
<td>$f_{CH_4}$</td>
<td>1.38 ± 0.05</td>
<td>(16)</td>
<td></td>
</tr>
<tr>
<td>$d[ln(CH_4 lifetime)]/d ln(CH_4)$</td>
<td>1.30</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>1.52 UCI</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>1.40 ± 0.10</td>
<td>Adopted</td>
<td></td>
</tr>
<tr>
<td>$d[O_3]/d[CH_4]$</td>
<td>3.7</td>
<td>± 0.8</td>
<td>(16)</td>
</tr>
<tr>
<td>DU($O_3$) ppm($CH_4$)$_{-1}$</td>
<td>2.4</td>
<td></td>
<td>(17)</td>
</tr>
<tr>
<td>$dF/d[CH_4]$</td>
<td>3.8</td>
<td>± 1.0</td>
<td>Adopted</td>
</tr>
<tr>
<td>mW m$^{-2}$ ppm($CH_4$)$_{-1}$</td>
<td>370</td>
<td>± 7.2%</td>
<td>(27)</td>
</tr>
<tr>
<td>$dF/d[O_3]$</td>
<td>48$^\dagger$</td>
<td></td>
<td>(29)</td>
</tr>
<tr>
<td>mW m$^{-2}$ DU($O_3$)$_{-1}$</td>
<td>48$^\dagger$</td>
<td></td>
<td>(29)</td>
</tr>
<tr>
<td>$dF/d[O_3]$</td>
<td>42$^\dagger$</td>
<td>± 12</td>
<td>20</td>
</tr>
<tr>
<td></td>
<td>36</td>
<td>± 6</td>
<td>(7)</td>
</tr>
<tr>
<td></td>
<td>36</td>
<td>± 8</td>
<td>Adopted</td>
</tr>
</tbody>
</table>

DU, Dobson unit.

*aStated uncertainties are one standard deviation (68% confidence interval).

†UCl values are from the UCI CTM used in this work. Adopted values are used throughout this analysis.

‡Model standard deviation is <25% without the single outlier model.

§RF calculated for preindustrial to present changes in tropospheric O$_3$, not aviation-specific pattern.

Table 2. Steady-state RF (mW m$^{-2}$) from aviation NOx emissions, 1 Tg(N) a$^{-1}$

<table>
<thead>
<tr>
<th>RF term</th>
<th>Factor</th>
<th>Model decomposition*</th>
<th>Source*</th>
</tr>
</thead>
<tbody>
<tr>
<td>Short-lived O$_3$</td>
<td>$+21.6 ± 7.2$</td>
<td>+27.3 ± 9.7</td>
<td></td>
</tr>
<tr>
<td>Long-lived CH$_4$</td>
<td>$−15.7 ± 3.6$</td>
<td>−16.1 ± 5.6</td>
<td></td>
</tr>
<tr>
<td>Long-lived O$_3$</td>
<td>$−5.3 ± 2.2$</td>
<td>−6.6 ± 3.3</td>
<td></td>
</tr>
<tr>
<td>Net</td>
<td>$+0.6 ± 8.3^*$</td>
<td>+4.5 ± 4.5</td>
<td></td>
</tr>
</tbody>
</table>

*Calculated from Eqs. 1–4 with factor values given in Table 1, assuming that all factors are uncorrelated. Stated uncertainties are one standard deviation (68% confidence interval).

†Calculated from models shown in Figs. 1 and 2. Stated uncertainties are one standard deviation.

§Accounting for correlations due to all common factors in Eqs. 2–4 reduces the uncertainty to ±8.1 mW m$^{-2}$.

F$_{short}$O$_3$. For the long-lived CH$_4$ response, however, the uncertainty in $dL_{CH_4}$/dE causes about 80% of the uncertainty in $F_{long}$CH$_4$. In both of these cases, the dominant uncertainty originates in factors that require aviation-specific studies. In contrast, for the long-lived O$_3$ response, almost 50% of the variance in $F_{long}$O$_3$ originates in the $d[O_3]/d[CH_4]$ term, which reflects uncertainty in the global response of O$_3$ to CH$_4$ perturbations and is not a unique response to aviation emissions. The $dL_{CH_4}$/dE and $dF/d[O_3]$ uncertainties explain most of the remaining variance of long-lived O$_3$ RF.

Correlation of RF Components and its Causes

Fig. 2 shows the ensemble of net RF across the models. For three early studies that did not calculate $F_{long}$O$_3$, we fill in values by scaling the reported $F_{long}$CH$_4$:

![Fig. 2. Steady-state net RF (mW m$^{-2}$) caused by aviation NOx emissions. Values are scaled to 1 Tg(N) a$^{-1}$. See Fig. 1 caption for the key to multimodel papers. An asterisk (*) indicates that the long-lived O$_3$ RF was obtained using Eq. 5. The mean and standard deviation is given.](image)
The net RF from aviation NOx for the model ensemble is $4.5 \pm 4.5$ mW m$^{-2}$ for emissions of 1 Tg(N) a$^{-1}$. Although 17 of the 21 net RF estimates are positive, the 68% confidence interval includes zero. This range is similar to those in recent studies that use a subset of the models analyzed here [e.g., $+4.3 \pm 3.4$ (20), $-0.3$ to $+6.0$ (31)].

The net RF derived from the factor decomposition is $+0.6$ mW m$^{-2}$. If we naively assume no correlation among the RF components, then its one-sigma range is $+0.6 \pm 3.8$ mW m$^{-2}$, which encompasses the ensemble-based estimate and likewise spans zero. However, this assumption of no correlation gives an uncertainty in the net RF that is nearly twice as large as the spread in models, despite the similarity of standard deviations for each RF component found above. This discrepancy persists if we account for correlations due to common factors in Eqs. 2-4 (uncertainty reduces to $\pm 8.1$ mW m$^{-2}$). The ensemble-derived standard deviation of net RF is also much smaller than would be expected if the variance of each of the ensemble-derived component RFs were independent (expected $\pm 11.6$ mW m$^{-2}$), implying that there are important correlations between the components.

Fig. 3 shows that short-lived O$_3$ RF and long-lived CH$_4$ RF are strongly correlated in the model ensemble ($R^2 = 0.79$), which has previously been assumed but not demonstrated (15). In the absence of quantitative data, earlier work assumed 100% anticorrelation, instead of the actual 79%, between aviation NOx RF that are too small (22). For example, if we naively assumed 100% anticorrelation, instead of the actual 79%, between the positive and negative RF components from the model ensemble, the inferred uncertainty in net RF would be $\pm 0.8$ mW m$^{-2}$, which is one-fifth of the actual ensemble variation. Assuming the same with the factor decomposition, we would expect net RF uncertainty of $\pm 1.4$ mW m$^{-2}$, which is one-third of the ensemble variation. Using a subset of the model ensemble studied here, Myhre et al. (7) found that the largest net aviation NOx RF occurred in models with the largest ratio of initial O$_3$ column change to fractional CH$_4$ lifetime change and the full ensemble also supports this conclusion.

To assess the causes of correlation between the RF components, we test the sensitivity of the University of California, Irvine (UCI) CTM to processes and parameters that likely reflect important differences among the models in the ensemble, similar to an earlier study of global O$_3$ and CH$_4$ budgets (32). Hoor et al. (20) describe the base CTM configuration (version 5.6). RF changes in the sensitivity tests are overlaid on Fig. 3 based on factor changes given in Table S1. Large and similar changes in both short-lived O$_3$ and long-lived CH$_4$ RF components come from halving either convective fluxes or surface NOx emissions (originally 45 Tg(N) a$^{-1}$). Both increase the aviation RF due to short-lived O$_3$ and decrease the (negative) RF due to long-lived CH$_4$ by 3 to 8 mW m$^{-2}$ each. Halving the lightning NOx source (originally 4 Tg(N) a$^{-1}$) similarly shifts both aviation RF components in the same direction as the first perturbations but by less than 2 mW m$^{-2}$, possibly because most lightning occurs in the tropics far from the major flight routes. Considering the export efficiency of surface NOx to the free troposphere is 5−20% (33, 34), the changes to aviation RF per unit reactive nitrogen supply to the free troposphere agree within a factor of 3 in these sensitivity tests. In all of these tests the RF responses in the UCI CTM lie along the axis of variation for the model ensemble, indicating that varying treatments of convection, surface emissions, and lightning NOx are plausible causes of the ensemble spread and anticorrelation of aviation RFs. Prior studies show that O$_3$ and CH$_4$ RF responses to aviation NOx emissions are sensitive to the spatial variation in background NOx within a single model (8). We suggest that variation in mean tropospheric NOx between models can explain the spread of O$_3$ and CH$_4$ RFs in the ensemble.

Changes in aviation RF with temporal development of the UCI CTM (previous version 5.4) also lie along the major axis of ensemble variation, although it is unclear which changes to transport, emissions, wet deposition, and photolysis are responsible. In another sensitivity test we halve the surface CO emissions (originally 983 Tg(CO) a$^{-1}$). Short-lived O$_3$ RF drops by 8 mW m$^{-2}$ in response, because CO is an O$_3$ precursor, but the long-lived CH$_4$ RF changes very little, possibly because of the offsetting decrease in OH consumption by CO. To test the effect of kinetic uncertainty on the change of RF by a factor of 2, we varied the adjustable rate coefficients for phsyochemistry reactions to the upper limit of their uncertainties (35) in favor of O$_3$ production from NOx (+15% for HO$_2$ + NO $\rightarrow$ OH + NO$_2$, −10% for NO + O$_3$ $\rightarrow$ NO$_2$ + O$_2$, −15% for HO$_2$ + O$_3$ $\rightarrow$ OH + 2O$_3$). This adjustment induces small changes in both RF components. In both of these sensitivity tests the slopes of O$_3$ vs. CH$_4$ RF responses depart from the slope of the model ensemble, which suggests that variations in kinetics and CO emissions between models are not major causes of the spread in reported aviation RFs. In a final sensitivity test (not plotted), we use 2004 meteorology instead of 2005. The RF change was negligibly small for both components, indicating that interannual variability has little effect on aviation RF.

Conclusions

RF of aviation NOx emissions is often evaluated through a model ensemble approach where the standard deviation of the ensemble is used as a measure of the uncertainty. From published results, we show that one important metric, steady-state RF, has not changed much in value or uncertainty with model and emission developments over the last decade. Overall, published work suggests that the steady-state, net RF is $4.5 \pm 4.5$ mW m$^{-2}$ for aviation NOx emissions of 1 Tg(N) a$^{-1}$, which is consistent with recent studies using a subset of the models analyzed here. Other important RFs from aviation include CO$_2$, contrails, and induced cirrus. Current CO$_2$ forcing is about $+28$ mW m$^{-2}$ for 2005 (22), but not in steady state. Contrail and cirrus RFs are typically estimated to be positive $[+33$ mW m$^{-2}$ (22)] although some recent studies find larger positive and negative cirrus effects $[−140$ to

\[ F_{\text{long CH}_4} = F_{\text{long CH}_4}(d[O_3]/d[CH_4])(dF/d[O_3]/dF/d[CH_4]). \]
120 mW m$^{-2}$ (36, 37)]. These steady-state contrail and cirrus RF estimates should be increased by about 15% to compare with the 1 Tg N$^{-1}$ values used here.

The GWP allows for direct comparison of aviation emissions of NOx with those of CO. From the model ensemble here, the net RF from 1 Tg N$^{-1}$ aviation emission pulse integrated over 100 y is $+4.5$ mW m$^{-2}$ and that of 1 Tg of CO is $0.087$ mW m$^{-2}$ (derived from ref. 27). Thus, the ensemble-derived GWP of aviation NOx (as N) is $52 \pm 52$. If we had included the additional cooling attributed to CH$_4$-driven changes in stratospheric water vapor (7), then this number would be closer to zero, but the uncertainty would increase. Similar 100-year GWP ranges were derived by Fuglestvedt et al. (31) (7 to 71) and Myhre et al. (7) (21 to 67), although the latter included the stratospheric water vapor effect. Given aviation CO$_2$ : NOx emission ratios of 800:1 by mass of N (7, 22), the 100-year integrated RF from aviation NOx is about $7\% \pm 7\%$ of that from aviation CO$_2$.

Using a complementary approach, we factor the aviation NOx RF into its key terms and assess the best value and uncertainty of those terms, many of which reflect basic knowledge of atmospheric chemistry and RF and are not specific to aviation. The three component RF values and their uncertainties propagated from the factor decomposition, assuming no correlations among the factors, agree with the ensemble mean values and standard deviations. This second method identifies those terms largely the factors, agree with the ensemble mean values and standard deviations from the factor decomposition, assuming no correlations among RF into its key terms and assess the best value and uncertainty of CH$_4$ RFs whereas a nonaviation term dominates for long-lived CH$_4$ and NOx with those of CO. This second method identifies those terms largely the factors, agree with the ensemble mean values and standard deviations. This trend is consistent with the well-known decrease in NOx production as NOx increases (e.g., 38, 39) although most aviation studies find no clear evidence of chemical nonlinearities. Improved assessments of the current and future climate impacts from aviation NOx depend on more accurate simulations of tropospheric composition, including advances in modeling convection, lightning, and surface emissions, as well as better characterization of the NOx distribution. Comparisons with NOx measurements in the free troposphere should identify which models match the observations and if these predict similar aviation impacts. However, current NOx climatologies (40) are assembled from campaigns of opportunity and do not have the necessary statistics or coverage to adequately test the models. An observational NOx climatology that includes NOx (OH + HO$_2$) sources and samples a wide range of latitudes and photochemical regimes in both hemispheres at seasonal intervals would aid process studies in global atmospheric chemistry and their application to changing emissions and climate.

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