Re: Manuscript acp-2016-1100, “Global deposition of total reactive nitrogen oxides from 1996 to 2014 constrained with satellite observations of NO2 columns” by Jeffrey A. Geddes and Randall V. Martin

Dear Handling Editor,

Please find attached our full response to the reviewer comments. We respond to each reviewer comment individually, followed by the relevant changes to our manuscript.

A full copy of our manuscript with tracked changes is included.

Thank you for your time and consideration of our manuscript.

Jeffrey A. Geddes
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Response to Referee #1

We thank the reviewer very much for their constructive comments. We respond to each comment individually below, followed by changes to the manuscript.

Reviewer Comment: My first comment is that the present manuscript is missing some description and explanation of the top-down NOx emissions. First, measurements from three satellite instruments are used to estimate the long-term NOx emissions. Are there any instrumental differences among the three satellite products? If so, how do you reconcile them?

Author Response: The referee makes a good point bringing up the instrumental differences between satellite products (a similar concern was raised by another reviewer). The most important differences are the overpass times and the horizontal footprint of individual observations. First, the GOME, SCIAMACHY, and GOME-2 overpass times are roughly 10:30 a.m., 10:00 a.m., and 9:30 a.m. respectively. Second, their horizontal footprints are roughly 320 km x 40 km, 60 km x 30 km, and 80 km x 40 km respectively. We reconciled these three records by consulting published literature and by closely examining the overlapping time periods ourselves, and concluded that a consistent time series is achieved without requiring additional corrections. The reason for this is largely because the daily satellite observations were all gridded to a regular coarse grid of 2° x 2.5° latitude by longitude. Using a comparison with long-term ground-based MAX-DOAS observations, Irie et al. (2012) demonstrated that there is no inherent biases in either SCIAMACHY or GOME-2 that would preclude their combination into a single record. The work of van der A (2008) and Konovalov et al. (2010) show that a self-consistent record can be achieved by downgrading the spatial footprint of the higher resolving instruments (e.g. through smoothing or convolution) to that of the lowest resolving instrument. This is what we have achieved by gridding all the observations to 2° x 2.5°. The combination of these observations is also aided by the fact that the retrieval algorithm for obtaining tropospheric NO2 column density from all three instruments is the same (http://www.temis.nl/airpollution/no2.html).

We further examined our approach by inspecting the timeseries from individual 2° x 2.5° pixels over selected populated regions. These are shown in Figure D1.

Given the evidence from this figure, and consensus in the literature, we concluded that the instrumental differences between instruments are inconsequential to our analysis.

In response to the referee’s comment, we have modified our manuscript to include the additional citations and to elaborate on our reasoning for combining the satellite instrument records despite their instrumental differences:

“We calculate top-down surface NOx emissions from 1996 to 2014 using observations from GOME (1995-2003), SCIAMACHY (2002-2011) and GOME-2 (2007- ). The similar overpass time of these three instruments (from about 9:30 a.m. to 10:30 a.m. local time) facilitates their combination to provide consistent long-term coverage (van der A et al., 2008; Konovalov et al. 2010; Geddes et al., 2016; Hilboll et al., 2013). We achieve consistency across all three instruments despite their varying pixel sizes (320 km x 40 km, 60 km x 30 km, and 80 km x 40 km for GOME, SCIAMACHY, and GOME-2 respectively) by gridding the daily observations from each to a regular coarse grid of 2° x 2.5° latitude by longitude.”
Figure D1: Monthly mean tropospheric NO2 from GOME, SCIAMACHY, and GOME-2 showing consistent agreement during overlap between instruments.

**RC:** Second, the three satellite instruments have some overlapping time periods (e.g. 2007-2011 for SCIAMACY and GOME-2). How do you use them for emission estimates during these periods? While these may have been presented in another study, a brief description here will help readers to better understand the method.

**AR:** In the case of GOME and SCIAMACHY, their period of overlap is small (June 2002 – April 2003). We use GOME observations alone for 2002, and SCIAMACHY observations alone for 2003. During the overlap between SCIAMACHY and GOME-2 (2007-2011), we use the GOME-2 observations given its more frequent global coverage (roughly every day) compared to SCIAMACHY (roughly every 6 days). We therefore use SCIAMACHY from years 2003-2006, and GOME-2 from years 2007-2014.

In response to the referee’s comment, we have added the following to our manuscript:

“In our study, we use GOME observations for years 1996 to 2002, SCIAMACHY observations for years 2003 to 2006, and GOME-2 observations for years 2007 to 2014”.

**RC:** And third, I strongly suggest present more information and analyses on the global top-down NOx emissions and their trends during the focused period by adding more text and a figure. This is not clear at present, and will be really helpful to understand the trends in NOy deposition as presented in Figure 6 and 7.
AR: We thank the referee for their suggestion, and agree that including details on the top-down emissions would be useful. We believe a table would be the optimal way to share this information. We have also included anthropogenic HNO$_3$ fluxes (from shipping emissions) to our calculation of total emissions to demonstrate full mass balance with NO$_x$ deposition.

In response to this comment, we have added the following material to our manuscript:

“Table 1 shows the annual global top-down NOx emissions from our calculations. We derive global mean satellite-constrained NOx emissions from 1996-2014 of 55.6 ± 3.4 Tg N yr$^{-1}$. Our top-down global NOx emissions for 2001 of 52.3 Tg N are consistent with the mean from over 20 models used in the Coordinated Model Studies Activities of the Task Force on Hemispheric Transport of Air Pollution (HTAP) for the same year of 46.6 ± 7.8 Tg N (Vet et al. 2014).”

Table 1: Global top-down NOx emissions calculated using the finite mass balance inversion approach in combination with observations from GOME, SCIAMACHY, and GOME-2.

<table>
<thead>
<tr>
<th>Year</th>
<th>Global NOx Emissions (Tg N yr$^{-1}$)</th>
</tr>
</thead>
<tbody>
<tr>
<td>1996</td>
<td>60.1</td>
</tr>
<tr>
<td>1997</td>
<td>58.4</td>
</tr>
<tr>
<td>1998</td>
<td>59.2</td>
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<td>1999</td>
<td>59.6</td>
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<tr>
<td>2000</td>
<td>53.4</td>
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<td>2001</td>
<td>52.3</td>
</tr>
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<td>2002</td>
<td>55.1</td>
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<td>2003</td>
<td>50.1</td>
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<td>2004</td>
<td>51.5</td>
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<td>2005</td>
<td>51.2</td>
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<td>2006</td>
<td>50.0</td>
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<td>2007</td>
<td>54.7</td>
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<tr>
<td>2008</td>
<td>56.1</td>
</tr>
<tr>
<td>2009</td>
<td>55.9</td>
</tr>
<tr>
<td>2010</td>
<td>57.5</td>
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<tr>
<td>2011</td>
<td>58.9</td>
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<tr>
<td>2012</td>
<td>59.3</td>
</tr>
<tr>
<td>2013</td>
<td>58.5</td>
</tr>
<tr>
<td>2014</td>
<td>54.0</td>
</tr>
<tr>
<td>Mean</td>
<td>55.6 ± 3.4</td>
</tr>
</tbody>
</table>

$^a$ Includes anthropogenic HNO$_3$ flux of 2.3 ± 0.1 Tg N yr$^{-1}$.

RC: Monthly mean simulated NO2 columns are calculated using days with coincident satellite observations”. How do you select coincident days when using monthly-mean gridded satellite NO2 observations (Line 11)? And how do you sample the model simulation? Please clarify.
First, we grid the daily NO2 tropospheric column retrievals to a regular 2° x 2.5° grid, then calculate monthly means ourselves. Therefore, we are able to keep track of which 2° x 2.5° grid boxes have satellite observations that pass quality control on each individual day. From our simulations we output late morning mean vertical NO2 column profiles every day. Using this information, we calculate monthly mean tropospheric NO2 columns from the model simulation only on days that are coincidently sampled by successful satellite observations.

In response to the referee’s comment, we have clarified this in our manuscript in the following places:

“We achieve consistency across all three instruments despite their varying pixel sizes (320 km x 40 km, 60 km x 30 km, and 80 km x 40 km for GOME, SCIAMACHY, and GOME-2 respectively) by gridding the daily observations from each to a regular coarse grid of 2° x 2.5° latitude by longitude.”

“In all cases, monthly mean simulated NO2 columns are calculated using days with coincident satellite observations. The simulated NO2 vertical column is output daily for late morning.”

Do you mean you do not change the seasonality of NOx emissions in the model? Please clarify. And what is the NOx emission seasonality in the model? This is not described in the Appendix.

The referee is correct. We have not changed the seasonality of NOx emissions in the model. We assume that the scaling factor determined from coincidently sampled model and satellite NO2 tropospheric columns applies uniformly to the model prior emissions all year long regardless of whether successful satellite observations are available.

In response to the referee’s comment, we have modified our manuscript to clarify our approach:

“In all cases, monthly mean simulated NO2 columns are calculated using days with coincident satellite observations. The simulated NO2 vertical column is output daily for late morning. We calculate scaling factors for every month with available satellite observations, then calculate an annual mean scaling factor that is used to infer annual mean from the mean monthly top-down emissions. Our top-down emissions retain the same seasonality as the prior emissions to mitigate concerns about seasonally missing data (such as from snow or monsoonal clouds).”

We have also added the following details regarding the emission seasonality to the Appendix:

“Monthly scaling of NOx emissions are included in North America (based on the VISTAS inventory), Europe (based on the EMEP inventory), and Asia (based on the Zhang et al. (2009) inventory). Monthly scaling of EDGAR emissions is based on the seasonality from the Global Emission Inventory Activity (Benkovitz et al 1996).”

Is there any trends in the export efficiency or changes in the export fraction during the period 1996-2014 over the US and Asia? From Line 11 below, it appears that the export fractions over Europe have a decreasing trend.

We thank the reviewer for their comment, and have evaluated the statistical significance of trends in export fraction over each region in the same manner as our evaluation of long term trends in deposition.
In response to the referee’s comment, we have added the following details to our manuscript:

“We estimate a similar fraction of NO\textsubscript{x} export from the continental US using our observationally-constrained simulation (34\% \pm 2\% from 1996-2014), with a small declining trend from a maximum of 38\% in 1999 to a minimum of 31\% in 2013.

“We calculate mean export of NO\textsubscript{x} emissions from western European countries to be 45\% \pm 4\%, with a notable decreasing trend from a maximum of 50\% in 1997 to a minimum of 39\% in 2014.”

“We estimate that an average of 24\% \pm 4\% emissions from China are exported, varying over time from as little as 15\% of emissions in 1998 to a maximum of 31\% of emissions in 2011 (an overall increasing trend).”

**RC:** A recent study on atmospheric nitrogen deposition over China reported a NO\textsubscript{y} export fraction of 36\% (Zhao et al., 2017), not that different from the values for Europe and the US, compared with 24\% in this study. Can you explain why? different NO\textsubscript{x} emissions, inclusion of adjacent oceans, or model horizontal resolution?

**AR:** We thank the reviewer for this important reference, which we have added to our manuscript. In our opinion, the most obvious explanation for the discrepancy is in model resolution or perhaps the rapidly changing satellite-constrained emissions over this period of time, which peak in China in 2011.

In response to the referee’s comment, we have made the following changes to our manuscript:

“We estimate that an average of 24\% \pm 4\% emissions from China are exported, varying over time from as little as 15\% of emissions in 1998 to a maximum of 31\% of emissions in 2011 (an overall increasing trend). Zhao et al. (2017) used a higher resolution (0.5° x 0.667°) GEOS-Chem simulation and estimated that 36\% of China’s NO\textsubscript{x} emissions over 2008-2012 are exported. We calculate an export fraction of around 27\% for the same time period. The discrepancy between the two estimates may be attributed to the coarser horizontal resolution of our simulation (2° x 2.5°), pointing to important resolution-dependent effects in global simulations of deposition. Other factors may include the use of different NO\textsubscript{x} emissions (our satellite-constrained emissions indicate rapid change over this period of time), and the treatment of adjacent oceans.”

**RC:** Please explain “perturbing NH\textsubscript{3} emissions everywhere”. Increase or decrease? Do you change all anthropogenic and natural NH\textsubscript{3} emissions, including the oceanic NH\textsubscript{3} emissions?

**AR:** In response to the referee’s comments, we have clarified the approach in our manuscript:

“Contemporary emissions of NH\textsubscript{3} are highly uncertain (Reis et al., 2009), so we perform a sensitivity experiment by perturbing (increasing) all anthropogenic and natural NH\textsubscript{3} emissions in the model by 25\% for the year 2012.”

**RC:** The unit “kg N ha\textsuperscript{-1} yr\textsuperscript{2}” here might be confusing. Suggest add here “at a rate of...” or use annual deposition changes during the period.
AR: We thank the referee for their suggestion. In response, we have modified our manuscript to use the following wording:

“NOy deposition declined most steeply throughout the northeastern United States at a rate of up to -0.6 kg N ha\(^{-1}\) yr\(^{-2}\).

“In Europe, statistically significant declines at a rate of up to -0.1 kg N ha\(^{-1}\) yr\(^{-2}\) are seen over some western countries.”

“On the other hand, NOy deposition has increased substantially throughout East Asia, exceeding a rate of +0.6 kg N ha\(^{-1}\) yr\(^{-2}\) in some parts.”


AR: We clarify this in our manuscript:

“For other species and for emissions beyond 2010, the closest available year is used.”

RC: Please state in the figure caption that the sensitivity test is for the year 2012.

AR: In response to the reviewer’s comment, we have added this to the figure caption.

REFERENCES:


Irie et al. (2012), Quantitative bias estimates for tropospheric NO2 columns retrieved from SCIAMACHY, OMI, and GOME-2 using a common standard for East Asia, *Atmospheric Measurement Techniques*, 5, 2403-2411

Konovalov et al. (2010) Multi-annual changes of NOx emissions in megacity regions: nonlinear trend analysis of satellite measurement based estimates, Atmos. Chem. Phys., 10, 8481–8498

van der A et al. (2008), Trends, seasonal variability, and dominant NOx source derived from a ten year record of NO2 measured from space, *Journal of Geophysical Research-Atmospheres*, 113, doi:10.1029/2007JD009021

Response to Referee #2

We thank the reviewer very much for their constructive comments. We respond to each comment individually below, followed by changes to the manuscript.

Reviewer Comment: My major comment is that I recommend a coherent section on model and data uncertainties that may affect your analysis and conclusions. Here are some examples of what such a discussion may include:

MERRA meteorological fields: Are there any biases in precipitation or transport that may affect your results, such as through simulated wet deposition? Are there any known biases that change over time in MERRA, such as occur as new observations are brought into the assimilation system over your 20 year simulation period? These are important biases to discuss as global coverage of surface observations (e.g., wet deposition) are sparse over most of the globe.

GEOS-Chem: No model is perfect? Any known issues?

Chemistry: What are the known chemistry uncertainties in the relevant reaction mechanisms? You’ve answered this with your sensitivity test in Section 3.4.

Emissions: Are there biases? For instance, are the NEI NOx emissions biased?

Author Response: We agree with the referee that there could be a larger discussion on model and data uncertainties that may affect our analysis.

In response to the reviewer’s comment, we have added the following discussion section to our manuscript:

“3.5 Other Considerations

A number of other uncertainties are important in an inversion of satellite NO2 columns to calculate surface NOx emissions and simulate long-term NOy deposition. These can depend on, for example, the choice of inversion approach, errors in the satellite retrieval, and uncertainties in model processes (e.g. emissions, boundary layer mixing, chemical NOx sinks, meteorology, and dry deposition).

Cooper et al. (2017) found that the finite mass balance inversion approach used here can be improved upon by using an iterative method that performs with similar accuracy as a four-dimensional variational data assimilation. Multi-constituent data assimilation also shows considerable promise for constraints on surface NOx emissions (Myazaki et al. 2017). Satellite retrieval algorithms continue to develop with advances that will improve the accuracy of future estimates of satellite-constrained NOy deposition.

Uncertainties in model processes are also of interest. For example, uncertainties in the chemical sink of NOx alone (e.g. the rate of HNO3 formation, heterogeneous loss of N2O5 onto aerosol) can have a substantial impact on top-down emissions estimates (Stavrakou et al., 2013), suggesting more fundamental work in constraining these processes is required. Lin et al. (2010) found that top-down NOx emissions estimates over East Asia are sensitive to other model uncertainties including planetary boundary layer mixing scheme, lightning emissions, diurnal profile of emissions, and a-priori NOx, CO, and VOC emissions. Uncertainties in model meteorology are also important. For example, the MERRA precipitation fields used in our study are known to correlate weakly with observational datasets (Rienecker et al., 2011), but improvements can be expected from MERRA-2 due to the inclusion of...
gauge- and satellite-based precipitation corrections (Reichle et al., 2017). Finally, dry deposition schemes are also highly variable among models (Flechard et al., 2011; Hardacre et al., 2015), and future work in dry deposition evaluation should be a priority.

Nonetheless, despite these uncertainties we find a high degree of consistency between observations and our predictions in the long-term changes to deposition. Evidence continues to emerge about potential biases in bottom-up inventories (e.g. Travis et al. 2016), and our observational constraint on NOx emissions mitigates against such biases. We expect continued advancements in inversion approaches, satellite retrieval algorithms, and fundamental atmospheric chemistry processes will allow for increasingly accurate satellite-based constraints on deposition.”

**RC:** Section 3.4: How does the model simulation of ammonia compare to observations, such as from AIRS, and the very long record of SO2, such as from the same instruments that you use for NO2

**AR:** The reviewer raises a good question. GEOS-Chem ammonia simulations have been explored by other groups (e.g. Schierl et al. 2016), and suggest that the model can underestimate interannual variability and concentrations of ammonia. A new NH3 emission inventory is now available (Paulot et al., 2014), but this inventory is still only representative of a short period of time (2005-2008), thus potentially introducing similar errors in other years of our analysis anyway. Our perturbation experiment with NH3 emissions (Section 3.4) is an attempt to investigate the dependence of NOy deposition on NH3 emission uncertainties. We found that reasonable uncertainties/changes in NH3 emissions (~25%) are generally inconsequential to the spatial distribution of predicted NOy deposition over land. We did not feel the need to repeat such an experiment with SO2 emissions, since SO2 emissions tend to be less uncertain than NH3 emissions and will therefore have even less of an impact on the predicted spatial distribution of NOy deposition. We argue that an evaluation of GEOS-Chem simulated NH3 and SO2 is thus beyond the scope of our study, and would not have substantial bearing on the conclusions in our manuscript.

**RC:** It is no easy task to create an inter-consistent long-term data record using multiple satellite observations, so this topic deserves some discussion. What are the uncertainties and potential biases? For example, a priori vertical profiles change over time

**AR:** We agree with the reviewer that creating a long-term record across multiple satellite records is not necessarily a trivial task (a similar concern was raised by another reviewer). In our case, we concluded that a consistent time series can be achieved without requiring additional corrections. The reason for this is largely because the daily satellite observations were all gridded to a regular coarse grid of 2° x 2.5° latitude by longitude. Using a comparison with long-term ground-based MAX-DOAS observations, Irie et al. (2012) demonstrated that there is no inherent biases in either SCIAMACHY or GOME-2 that would preclude their combination into a single record. The work of van der A (2008) and Konovalov et al. (2010) show that a self-consistent record can be achieved by downgrading the spatial footprint of the higher resolving instruments (e.g. through smoothing or convolution) to that of the lowest resolving instrument. This is what we have achieved by gridding all the observations to 2° x 2.5°. The combination of these observations is also aided by the fact that the retrieval algorithm for obtaining tropospheric NO2 column density from all three instruments is the same.
(http://www.temis.nl/airpollution/no2.html), and by the fact that their overpass times are similar (between 9:30 a.m. and 10:30 a.m. local time).

We tested our approach by inspecting the timeseries from individual 2° x 2.5° pixels over selected populated regions. These are shown in Figure D1.

![Figure D1: Monthly mean tropospheric NO2 from GOME, SCIAMACHY, and GOME-2 show consistent agreement during overlap between instruments.](image)

Given this and the consensus in the literature, we conclude that the instrumental differences between each satellite instrument are inconsequential to our analysis.

In response to the referee’s comment, we have modified our manuscript to include additional citations, and to elaborate on our reasoning for combining the satellite instrument records despite their instrumental differences:

“We calculate top-down surface NOx emissions from 1996 to 2014 using observations from GOME (1995-2003), SCIAMACHY (2002-2011) and GOME-2 (2007- ). The similar overpass time of these three
instruments (from 9:30 a.m. to 10:30 a.m. local time) facilitates their combination to provide consistent long-term coverage (van der A et al., 2008; Konovalov et al. 2010; Geddes et al., 2016; Hilboll et al., 2013). We achieve consistency across all three instruments despite their varying pixel sizes (320 km x 40 km, 60 km x 30 km, and 80 km x 40 km for GOME, SCIAMACHY, and GOME-2 respectively) by gridding the daily observations from each to a regular coarse grid of 2° x 2.5° latitude by longitude.”

The referee also brings up the issue of changing a-priori vertical profiles over time. We agree that vertical profile shapes are an integral component of the tropospheric NO2 column retrieval, and indeed are a significant contributor to column error (at least ~10% according to Boersma et al. (2004), and up to ~100% according to Laughner et al. (2016)). However, we are limited by the current generation of available global tropospheric NO2 retrieval algorithms. Advances in satellite-derived tropospheric NO2 column retrievals will be integral to improvements in top-down emissions estimates. In response to the reviewer comment, we address this in modifications we have made to the manuscript:

“The error in individual satellite-derived tropospheric NO2 column retrievals to be around 35-60% for polluted scenes and greater than 100% for clean regions (Boersma et al. 2004).”

“We expect continued advancements in inversion approaches, satellite retrieval algorithms, and fundamental atmospheric chemistry processes will allow for increasingly accurate satellite-based constraints on deposition.”

RC: Page 4, Line 9: Since the topic of this Nowlan paper is similar and from the same group, it may be worth a sentence describing the major conclusion of this paper and how your manuscript is different/better. In fact, you may want to do briefly so the same for the other papers mentioned in this same paragraph.

AC: We thank the reviewer for this suggestion. We agree that it would be insightful to summarize the important points from Nowlan et al., although the statistical approaches taken by the other examples vary substantially from our approach and are less directly comparable. In response to this comment, we have included the following in our manuscript:

“Nowlan et al. (2014) demonstrated how satellite-inferred surface concentrations of NO2 can be combined with modeling to produce spatially continuous estimates of NO2 dry deposition fluxes. They found that dry deposition of NO2 contributes as much as 85% of total NOy deposition in urban areas, but only represents 3% of global NOx emitted. The remaining 97% of global NOy deposition is made up of both wet and dry deposition of other reactive nitrogen oxide compounds that are not directly observed by satellite-based instruments.”

RC: Figure 2. The two rows of plots look identical. Is there any way to show the differences between the two periods. If not, I’m not sure it’s helpful to show both rows.

AC: We are very grateful to the reviewer for their careful attention to this figure. In fact, the two panels are identical and this was an error in the original figure (both top and bottom accidentally show results for 2000-2002). We have corrected this mistake and sincerely aplogise for the error. A new version of this figure will be uploaded with the corrected manuscript, and is included here as Figure D2 for the reviewer’s reference. The differences between the two time periods are now more obvious. Declines in
wet nitrate deposition over this period are evident over North America and parts of Western Europe, while increases are evident over East Asia. Note that the observational coverage is also slightly different across the two periods.

Figure D2: Corrected version of Figure 2 from the manuscript, showing results from the 2000-2002 comparison in the top panels, and results from the 2005-2007 comparison in the bottom panels.

REFERENCES:


Irie et al. (2012), Quantitative bias estimates for tropospheric NO2 columns retrieved from SCIAMACHY, OMI, and GOME-2 using a common standard for East Asia, *Atmospheric Measurement Techniques*, 5, 2403-2411

Konovalov et al. (2010) Multi-annual changes of NOx emissions in megacity regions: nonlinear trend analysis of satellite measurement based estimates, Atmos. Chem. Phys., 10, 8481–8498

van der A et al. (2008), Trends, seasonal variability, and dominant NOx source derived from a ten year record of NO2 measured from space, *Journal of Geophysical Research-Atmospheres*, 113, doi:10.1029/2007JD009021


Lin et al. (2010), Constraint of anthropogenic NOx emissions in China from different sectors: a new methodology using multiple satellite retrievals, *Atmos. Chem. Phys.*, 10, 63–78
Stavrakou et al. (2013), Key chemical NOx sink uncertainties and how they influence top-down emissions of nitrogen oxides, Atmos Chem Phys, 13, 9057-9082


Reichle et al. (2017), Assessment of MERRA-2 land surface hydrology estimates, Journal of Climate, 30, DOI: 10.1175/JCLI-D-16-0720.1

Flechard et al. (2010), Dry deposition of reactive nitrogen to European ecosystems: a comparison of inferential models across the NitroEurope network, Atmos Chem Phys, 11, 2703-2728

Hardacre et al. (2015), An evaluation of ozone dry deposition in global scale chemistry climate models, Atmos Chem Phys, 15, 6419-6436

Schiferl et al. (2016), Interannual variability of ammonia concentrations over the United States: sources and implications, Atmos Chem Phys, 16, 12305-12328
Global deposition of total reactive nitrogen oxides from 1996 to 2014 constrained with satellite observations of NO₂ columns

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Abstract

Reactive nitrogen oxides (NOₓ) are a major constituent of the nitrogen deposited from the atmosphere, but observational constraints on their deposition are limited by poor or nonexistent measurement coverage in many parts of the world. Here we apply NO₂ observations from multiple satellite instruments (GOME, SCIAMACHY, and GOME-2) to constrain the global deposition of NOₓ over the last two decades. We accomplish this by producing top-down estimates of NOₓ emissions from inverse modeling of satellite NO₂ columns over 1996-2014, and including these emissions in the GEOS-Chem chemical transport model to simulate chemistry, transport, and deposition of NOₓ. Our estimates of long-term mean wet nitrate (NO₃⁻) deposition are highly consistent with available measurements in North America, Europe, and East Asia combined (r = 0.83, normalized mean bias = -7%, N = 136). Likewise, our calculated trends in wet NO₃⁻ deposition are largely consistent with the measurements, with 129 of the 136 gridded model-data pairs sharing overlapping 95% confidence intervals. We find that global mean NOₓ deposition over 1996-2014 is 56.0 Tg N yr⁻¹, with a minimum in 2006 of 50.5 Tg N and a maximum in 2012 of 60.8 Tg N. Regional trends are large, with opposing signs in different parts of the world. Over 1996 to 2014, NOₓ deposition decreased by up to 60% in eastern North America, doubled in
regions of East Asia, and declined by 20% in parts of Western Europe. About 40% of the
global NO$_x$ deposition occurs over oceans, with deposition to the North Atlantic Ocean
decreasing and deposition to the northwestern Pacific Ocean increasing. Using the residual
between NO$_x$ emissions and NO$_y$ deposition over specific land regions, we investigate how
NO$_x$ export via atmospheric transport has changed over the last two decades. Net export from
the continental United States decreased substantially, from 2.9 Tg N yr$^{-1}$ in 1996 to 1.5 Tg N
yr$^{-1}$ in 2014. On the other hand, export from China more than tripled between 1996 and 2011
(from 1.0 Tg N yr$^{-1}$ to 3.5 Tg N yr$^{-1}$), before a striking decline to 2.5 Tg N yr$^{-1}$ by 2014. We
find that declines in NO$_x$ export from some Western European countries have counteracted
increases in emissions from neighbouring countries to the east. A sensitivity study indicates
that simulated NO$_y$ deposition is robust to uncertainties in NH$_3$ emissions with a few
exceptions. Our novel long-term study provides timely context on the rapid redistribution of
atmospheric nitrogen transport and subsequent deposition to ecosystems around the world.

1 Introduction

The introduction of reactive nitrogen to the environment by anthropogenic activities (e.g.
from fossil fuel combustion and the production of fertilizers for agriculture) has drastically
altered the global nitrogen cycle with consequences throughout the Earth system (Galloway
2004). Reactive nitrogen dominates the chemical production of tropospheric ozone and
contributes to inorganic aerosol formation, with implications for air quality and climate.
Deposition of nitrogen from the atmosphere has been linked to eutrophication and
acidification (Bouwman et al., 2002), reductions in biodiversity (Bobbink et al., 2010;
Hernández et al., 2016; Isbell et al., 2013; De Schrijver et al., 2011), and climate impacts
through coupling with the carbon cycle and greenhouse gas emissions (Liu and Greaver,
2009; Reay et al., 2008; Templer et al., 2012). Despite its global importance, observational
constraints on nitrogen deposition are lacking in many parts of the world due to poor or
nonexistent measurement coverage (Vet et al., 2014).

Atmospheric transport is a dominant process for distributing reactive nitrogen around the
world (Galloway et al., 2008). Some forms of reactive nitrogen can be transported over
distances greater than 1000 km (Neuman et al., 2006; Sanderson et al., 2008), depositing
across national boundaries and continents. For example, the U.S. is estimated to export 30-
40% of its reactive nitrogen emissions (Dentener et al., 2006; Holland et al., 2005; Zhang et
while transport from China accounts for up to 66-92% of total nitrogen deposition to parts of the northwestern Pacific Ocean (Zhao et al., 2015). Fertilization of the open ocean due to atmospheric transport and deposition of anthropogenic nitrogen may be a considerable factor in marine productivity (Duce et al., 2008), prompting important questions about the fate and impact of deposition far downwind of sources where observations are limited.

Reactive nitrogen oxides (NO\(_x\) ≡ NO + NO\(_2\) + HNO\(_3\) + HONO + organic nitrate molecules + aerosol nitrate) contribute about half of the total nitrogen deposited worldwide (Dentener et al., 2006). NO\(_y\) deposition was estimated to be around 45-50 Tg N yr\(^{-1}\) in the late-1990s and early 2000s, representing a 3-4 fold increase since the pre-industrial era (Dentener et al., 2006; Kanakidou et al., 2016; Lamarque et al., 2013). A substantial range exists in the trajectory of global NO\(_y\) deposition beyond the year 2000 depending on emission scenario. Galloway et al. (2004) projected that NO\(_y\) deposition could increase by >70% by 2050, while Dentener et al. (2006) projected changes between -25% to +50% by 2030 for maximum feasible reduction and “pessimistic” scenarios respectively. More recent multi-model projections by Lamarque et al. (2013) estimate NO\(_y\) deposition would change by -16% to +5% for 2030 and by -48% to -25% for 2100, depending on the Representative Concentration Pathway (RCP) scenario. This range in projections highlights the need for the global observational constraints on contemporary changes in nitrogen oxide emissions.

Sources of NO\(_x\) (≡ NO + NO\(_2\)), whose oxidation is responsible for the formation of other reactive nitrogen oxides, include fossil fuel combustion, biomass burning, lightning, and biogenic emission from soil. The magnitude and spatial distribution of NO\(_x\) emissions have changed considerably over the past several decades, corresponding to patterns of human development and emission control measures. Tropospheric NO\(_2\) columns derived from satellite remote sensing observations have been used extensively to constrain regional and global NO\(_x\) emissions (Streets et al., 2013). This top-down approach complements bottom-up inventories that are assembled using regionally specific emission factors and fuel combustion data for various source categories. In particular, satellite NO\(_2\) observations can provide insight into otherwise poorly constrained sources (Beirle et al., 2010; Jaegle et al., 2005; Richter et al., 2004; Vinken et al., 2014), produce coherent long-term trends (van der A et al., 2008; Lu et al., 2015; Stavrakou et al., 2008; Zhang et al., 2007), document interannual variability (Castellanos and Boersma, 2012; Konovalov et al., 2010a; Russell et al., 2012), offer information to evaluate and improve bottom-up inventories at the global scale (Martin, 2003;
Satellite observations of NO$_2$ began with GOME (1995-2003) followed by SCIAMACHY (2002-2011), and continue today with OMI (2004- ), GOME-2 (2007- ), and TROPOMI (2017- ), resulting in a record of global atmospheric NO$_2$ abundance over the past 20 years. These observations have been used previously to estimate the deposition of nitrogen species, either in combination with chemical transport modeling or with empirical approaches (Cheng et al., 2013; Jia et al., 2016; Lu et al., 2013; Nowlan et al., 2014). For example, Nowlan et al. (2014) demonstrated how satellite-inferred surface concentrations of NO$_2$ can be combined with modeling to produce spatially continuous estimates of NO$_2$ dry deposition fluxes. They found that dry deposition of NO$_2$ contributes as much as 85% of total NO$_x$ deposition in urban areas, but represents only 3% of global NO$_x$ emitted. The remaining 97% of global NO$_x$ deposition is made up of both wet and dry deposition of other reactive nitrogen oxide compounds that are not directly observed by satellite-based instruments.

In this study, we expand on the approach of Nowlan et al. (2014) by using the satellite observations of NO$_2$ columns to constrain total NO$_y$ deposition, including other oxidized nitrogen species and wet deposition which contribute substantially to NO$_y$ deposition. We accomplish this by constraining surface NO$_x$ emissions using the satellite observations of NO$_2$, and simulating subsequent NO$_y$ deposition with a global chemical transport model. Given the effective mass balance between NO$_x$ emissions and deposition of reactive nitrogen oxides, observational constraints on NO$_x$ emissions provide a powerful top-down constraint on deposition (which to our knowledge has not yet been exploited in this way).

We leverage the long-term coverage of GOME, SCIAMACHY, and GOME-2 observations to produce a globally consistent and continuous record of NO$_y$ deposition from 1996 to 2014. We highlight long-term trends in satellite-constrained NO$_y$ deposition around the world and discuss changes in regional export of NO$_x$. Our satellite-constrained estimates of NO$_y$ deposition are evaluated using measured wet nitrate (NO$_3^-$) deposition from a variety of sources worldwide. We also explore the sensitivity of the NO$_y$ deposition estimates to uncertainties in NH$_3$ emissions.
2 Methods

2.1 Satellite-based constraints on NO\textsubscript{y} deposition

The application of satellite NO\textsubscript{2} column observations to constrain NO\textsubscript{y} deposition requires a method to propagate the observational information across the entire NO\textsubscript{y} system containing species with lifetimes of days or longer. The short NO\textsubscript{x} lifetime of hours during daytime satellite observations implies that a direct assimilation for NO\textsubscript{2} column abundance would rapidly lose the observational information as the assimilation returns to its unperturbed state well before the next satellite observation days later. We therefore apply satellite NO\textsubscript{2} observations to constrain NO\textsubscript{x} emissions in a simulation of NO\textsubscript{y} deposition so the information propagates across the entire NO\textsubscript{y} system.

We calculate top-down surface NO\textsubscript{x} emissions from 1996 to 2014 using observations from GOME (1995-2003), SCIAMACHY (2002-2011) and GOME-2 (2007-). The similar overpass time of these three instruments (from about 9:30 a.m. to 10:30 a.m. local time) facilitates their combination to provide consistent long-term coverage (Geddes et al., 2015; Hilboll et al., 2013; Konovalov et al. 2010; van der A et al., 2008). We achieve consistency across all three instruments despite their varying pixel sizes (320 km x 40 km, 60 km x 30 km, and 80 km x 40 km for GOME, SCIAMACHY, and GOME-2 respectively) by gridding the daily observations from each to a regular coarse grid of 2° x 2.5° latitude by longitude.

Tropospheric NO\textsubscript{2} vertical column densities are provided by KNMI at http://www.temis.nl/airpollution/. In all cases, NO\textsubscript{2} column densities are retrieved by differential optical absorption spectroscopy using backscattered radiance in the 425-450 nm wavelength range according to the retrieval algorithm documented in Boersma et al. (2004). The error in individual satellite-derived tropospheric NO\textsubscript{2} column retrievals to be around 35-60% for polluted scenes and greater than 100% for clean regions (Boersma et al. 2004). Boersma et al. (2016) well describe the value of accounting for vertically-resolved instrument sensitivity. We use the averaging kernels provided with the data to replace a priori NO\textsubscript{2} vertical profiles with those from GEOS-Chem model following Lamsal et al. (2010). We use daily nadir observations with a cloud radiance fraction of less than 0.5. We minimize errors associated with wintertime retrievals by using a solar zenith angle cut-off of 50°.

We use the GEOS-Chem chemical transport model (www.geos-chem.org) v9-02 to conduct the inversion of satellite observations of NO\textsubscript{2} and constrain global NO\textsubscript{y} deposition.
The simulation is described in Appendix 1. Briefly, GEOS-Chem is driven by assimilated meteorology from the NASA Global Modeling and Assimilation Office and simulates detailed HO\textsubscript{x}-NO\textsubscript{x}-VOC-aerosol chemistry (Bey et al., 2001; Park et al., 2004). Removal occurs through wet deposition (Amos et al., 2012; Liu et al., 2001), and dry deposition based on the widely used resistance-in-series formulation (Wesely, 1989). Anthropogenic, biogenic, soil, lightning, and biomass burning emissions are included (see Appendix 1). In the case of NO\textsubscript{x}, the bottom-up emissions are used as prior estimates that we then overwrite with the top-down emissions.

We adopt a finite-difference mass balance inversion (Cooper et al., 2017; Lamsal et al., 2011) for computational expedience given the 19-year period of interest. In two initial simulations, a perturbation (30%) to the a-priori emissions, \(E\), in a grid cell is used to find the relationship between the a-priori NO\textsubscript{2} column, \(\Omega\), and the change in the column resulting from that perturbation:

\[
\frac{\Delta E}{E} = \beta \times \frac{\Delta \Omega}{\Omega}
\]  

The factor \(\beta\) in Equation 1 accounts for non-linear feedbacks between a change in NO\textsubscript{x} emissions and NO\textsubscript{x} chemistry in a grid cell leading to grid cell NO\textsubscript{2} column abundance.

We then use monthly-mean gridded satellite observations, \(\Omega_{sat}\), in combination with monthly \(\beta\) values for each grid box to derive new gridded annual emissions, \(E_{topdown}\), from the prior emissions estimates, \(E_{prior}\), by rewriting Equation 1 as:

\[
E_{topdown} = E_{prior} \cdot \left[1 - \beta \frac{\Omega_{sat} - \Omega_{prior}}{\Omega_{prior}}\right]
\]  

We use GOME observations for years 1996 to 2002, SCIAMACHY observations for years 2003 to 2006, and GOME-2 observations for years 2007 to 2014. In all cases, monthly mean simulated NO\textsubscript{2} columns are calculated using days with coincident satellite observations. The simulated NO\textsubscript{2} vertical column is output daily for late morning. We calculate annual mean scaling factors for every month with available satellite observations, then calculate an annual mean scaling factor that is used to infer annual mean from the mean monthly top-down emissions. Our top-down emissions retain the same seasonality as the prior emissions to mitigate concerns about seasonally missing data (such as from snow or monsoonal clouds). The top-down emissions are then used in a final simulation to model NO\textsubscript{y} deposition. For
locations without any satellite observations, the a-priori emissions are used. The resultant simulation of NO\textsubscript{x} deposition is thus constrained by, and consistent with, the satellite NO\textsubscript{2} observations (similar in essence to an assimilation system). We note uncertainty in tropospheric NO\textsubscript{2} from lightning will propagate into the inversion (Travis et al., 2016)\textsuperscript{(Travis et al. 2016)}, but there is no evidence of a significant trend in lightning NO\textsubscript{x} over the long term (Murray et al. 2012). A constant bias is unlikely to affect the trend analyses presented here.

Table 1 shows the annual global top-down NO\textsubscript{x} emissions we obtain from our calculations. We derive global mean satellite-constrained NO\textsubscript{x} emissions from 1996-2014 of 553.62 ± 3.4 Tg N yr\textsuperscript{-1}. Our top-down global NO\textsubscript{x} emissions for 2001 of 529.39 Tg N are consistent with the mean ± standard deviation from over 20 models used in the Coordinated Model Studies Activities of the Task Force on Hemispheric Transport of Air Pollution (HTAP) for the same year of 46.6 ± 7.8 Tg N (Vet et al. 2014).

### 2.2 Measurements of Wet Deposition

We use a variety of regional and global measurements of wet nitrate (NO\textsubscript{3}\textsuperscript{-}) deposition to evaluate our satellite-constrained simulation from 1996 to 2014.


To evaluate the long-term means and trends from 1996-2014, we obtain observations of wet NO\textsubscript{3}\textsuperscript{-} deposition from North America, Europe, and East Asia where continuous measurements are available throughout most of our study period. Observations come from the National Atmospheric Deposition Program in the United States (http://nadp.sws.uiuc.edu/, available 1996-2014), from the Canadian Air and Precipitation Monitoring Network in Canada (http://www.ec.gc.ca/rs-mn/, available 1996-2011), from the European Monitoring and Evaluation Programme in Europe (http://www.emea.int/, available 1996-2014) and from the Acid Deposition Monitoring Network in East Asia (http://www.eanet.asia, available 2000-2014). In the US and Canada, wet deposition is measured by wet-only samplers that are triggered at the onset of precipitation. Measurements in Europe are made by bulk- and wet-only sampling methods and we used both in this analysis. Measurements across East Asia are
reported as wet-only, although at some stations this may not be accomplished by strictly wet-

For our analysis, we only included stations which had quality controlled annual data for
at least 15 of the 19 years in our study. This left 128 stations across the United States, 14
stations in Canada, 18 stations across Europe, and 14 stations across East Asia. For
comparison with the GEOS-Chem model, if multiple stations are available within a single
grid box we grid all measurements of annual wet deposition to the model horizontal
resolution.

3 Satellite-Constrained Estimates of NO\textsubscript{y} Deposition

Here we summarize the overall patterns in long-term mean deposition resulting from our
satellite-constrained simulation, followed by a discussion of the long term trends, changes in
regional export, and the sensitivity of the simulated NO\textsubscript{y} deposition to potential uncertainties
in NH\textsubscript{3} emissions.

3.1 Long-term Mean NO\textsubscript{y} Deposition

Figure 1 (top) shows our satellite-constrained long-term mean NO\textsubscript{y} deposition from
1996 to 2014. We find that 32.2 Tg N yr\textsuperscript{-1} is deposited on average over the continents (57% of
the total), and 23.8 Tg N yr\textsuperscript{-1} is deposited on average over the oceans (43% of the total). This
is similar to the estimate by Galloway et al. (2004) that 46% of modern day NO\textsubscript{y} deposition
occurs over the oceans. Critical nitrogen deposition loads for various natural freshwater and
terrestrial ecosystems lie in the range of 5-20 kg N ha\textsuperscript{-1} yr\textsuperscript{-1}, depending on the ecosystem, soil
conditions, and land history (World Health Organization, 2000). We estimate that mean
deposition of oxidized nitrogen alone exceeds 5 kg N ha\textsuperscript{-1} yr\textsuperscript{-1} over a land area of
approximately 12.7 x 10\textsuperscript{6} km\textsuperscript{2} (or ~8% total land area).

In the Northern Hemisphere, high NO\textsubscript{y} deposition tends to be associated with regions
that have high anthropogenic NO\textsubscript{x} sources. We find mean NO\textsubscript{y} deposition in the eastern
United States exceeds 10 kg N ha\textsuperscript{-1} yr\textsuperscript{-1} (maximum = 11.4 kg N ha\textsuperscript{-1} yr\textsuperscript{-1}) with elevated
deposition extending into southeastern Canada and hundreds of kilometers into the Atlantic
Ocean. This is similar to the multi-model ensemble results from ACCMIP and HTAP,
predicting between 5-15 kg N ha\textsuperscript{-1} yr\textsuperscript{-1} in this region (Lamarque et al. 2013; Vet et al. 2014).
A prior GEOS-Chem analysis over North America for the years 2006-2008 also predicted NO$_y$ deposition exceeding 10 kg N ha$^{-1}$ yr$^{-1}$ in the eastern US (Zhang et al. 2012). Elsewhere in North America, we find high NO$_y$ deposition along the west coast of California (up to 6 kg N ha$^{-1}$ yr$^{-1}$) and in the vicinity of Mexico City (up to 10 kg N ha$^{-1}$ yr$^{-1}$).

We find mean NO$_y$ deposition is also elevated throughout Europe, with a maximum of 8.5 kg N ha$^{-1}$ yr$^{-1}$ located in northern Italy near the Po Valley region. Again, our long-term estimate in this region is similar to the ACCMIP and HTAP ensemble means, predicting NO$_y$ deposition in the range of 5-10 kg N ha$^{-1}$ yr$^{-1}$ (Lamarque et al. 2013; Vet et al. 2014). The elevated deposition here is also spatially consistent with the results from Holland et al. (2005).

We find high deposition extending into western Russia with a hotspot in the vicinity of Moscow approaching 5 kg N ha$^{-1}$ yr$^{-1}$. Our observation-constrained estimate also has isolated regions of high deposition in the Middle East (around 4-5 kg N ha$^{-1}$ yr$^{-1}$ in the vicinity of Tehran and around the Persian Gulf).

We find that the highest mean deposition in the world occurs in China, exceeding 10 kg N ha$^{-1}$ yr$^{-1}$ in many regions. High deposition extends into the mid-latitude western Pacific Ocean off the coast of East Asia. NO$_y$ deposition in the ACCMIP and HTAP ensemble means also exceeds 10 kg N ha$^{-1}$ yr$^{-1}$ throughout eastern China. We find the highest long-term mean deposition (with a maximum close to 20 kg N ha$^{-1}$ yr$^{-1}$) occurs in the south, around the Pearl River Delta and in the vicinity of Guangzhou, although deposition is also high in the regions just west of Beijing and Shanghai.

In the Southern Hemisphere, high NO$_y$ deposition is associated with biomass burning and soil NO$_x$ sources, in addition to anthropogenic sources. For example, we find NO$_x$ deposition is between 3 to 5 kg N ha$^{-1}$ yr$^{-1}$ in central and southern Brazil, and in the tropical rainforests and moist savannahs of Africa. Our estimates in these biomass burning and soil NO$_x$ dominated regions are also generally consistent with the ACCMIP and HTAP ensemble estimates (2-5 kg N ha$^{-1}$ yr$^{-1}$). We find NO$_y$ deposition up to 10 kg N ha$^{-1}$ yr$^{-1}$ in the vicinity of Sao Paulo and Rio de Janeiro, and in the vicinity of Johannesburg and the industrialized Mpumalanga Highveld of South Africa (all dominated by anthropogenic NO$_x$ emissions). Our constrained simulation also identifies hotspots of deposition in the vicinity of Melbourne and Sydney, Australia (~4 kg N ha$^{-1}$ yr$^{-1}$).

Figure 1 (bottom) shows the simulated long-term ratio of dry NO$_y$ deposition to total (wet + dry) NO$_y$ deposition. Globally, dry and wet deposition contribute roughly equally to
total NO$_y$ deposition (52% and 48% respectively). Dry deposition usually accounts for more than 50% of the total over the continents and directly off shore whereas wet deposition dominates over the remote oceans. In the generally arid regions of the world (e.g. southwestern US, the Sahara Desert, the Arabian Peninsula, and the Gobi Desert) dry deposition accounts for ~85% or more of the total deposition. Elsewhere, dry deposition fractions tend to be highest (>60%) nearest to major surface NO$_x$ sources (e.g. eastern US, Western Europe, and near other major urban centres around the world in addition to the soil and biomass-burning dominated source regions in South America and Africa). HNO$_3$ typically makes the dominant contribution to dry NO$_y$ deposition, although NO$_2$ and HNO$_3$ can make almost equal contributions in certain high-NO$_x$ environments. Isoprene nitrates and peroxyacetyl nitrates comprise ~10-30% of dry NO$_y$ deposition in some densely forested and high latitude environments respectively.

We evaluate our estimates of NO$_y$ deposition with measured wet NO$_3^-$ from several sources. Figure 2 shows measurements of annual wet NO$_3^-$ deposition from the World Data Centre for Precipitation Chemistry, available for two time periods: 2000-2002 (N = 470) and 2005-2007 (N = 484). In both we see the patterns of elevated deposition in eastern North America, Western Europe, and parts of South and East Asia, with lower deposition in western North America, across high latitudes in the Northern Hemisphere, and in the available observations in Africa. High deposition in the Southern Hemisphere is observed between Sao Paulo and Rio de Janeiro, and just southeast of Johannesburg. Figure 2 also shows the wet NO$_3^-$ deposition from our constrained simulation during the same two time periods (2000-2002 and 2005-2007), which exhibits similar patterns found in total NO$_y$ deposition (Figure 1).

We find a high degree of consistency between our estimate and the observations for both the 2000-2002 (N = 306 model-data pairs) and 2005-2007 (N = 310 model-data pairs). The normalized mean bias (NMB) is -14% and -16% respectively. The vast majority of pairs (> 80%) agree to within 50% of each other. Figure 3 shows scatter plots for specific subsets of the global data. The agreement for both time periods is strongest over North America (r = 0.92 for both 2000-2002 and 2005-2007, NMB = +1.0% and -5.0% respectively). Robust model agreement with wet nitrate deposition observations over densely monitored North America is characteristic of other global studies (Dentener et al. 2006; Lamarque et al. 2013; Vet et al. 2014). Our agreement is also good in Europe (r = 0.69 and
0.66, and NMB = -31.0% and -29.8% respectively). The weaker correlation and low bias in this region is likewise characteristic of global studies, although our spatial correlation (r=0.66-0.69) is on the high end of previously reported multi-model ensembles (r ~ 0.4-0.6, Dentener et al. 2006; Lamarque et al. 2013; Vet et al. 2014). The negative bias over Europe compared to North America has previously been attributed to poor modeling of precipitation, and/or spatial representativeness of the measurements compared to model resolution. Throughout the rest of the world (encompassing observations mostly over Asia, but also over eastern Russia, and some locations in the Southern Hemisphere) the combined spatial coverage of the observations is very low (N = 53). Normalized mean bias in these estimates is also high compared to North America (NMB = -19.5% and -17.8% for 2000-2002 and 2005-2007 respectively), and our spatial correlation with the measurements is weak (r = 0.35 and 0.42 respectively). We find that our poor agreement here is disproportionately driven by the two observations that also have the highest measured deposition in the world: near Port Blair on the South Andaman Island in the Bay of Bengal, and in the Arunachal Pradesh state in northeastern India. Agreement is considerably better with the rest of the data (r = 0.78 and 0.72, NMB = +0.01% and -0.01% for 2000-2002 and 2005-2007 respectively). Excluding these two points substantially improves the global agreement as well (from r = 0.57 to 0.75 and r = 0.59 to 0.75 respectively). Site representativeness, precipitation errors, or uncertainty in our satellite-constrained NOX emissions may explain the discrepancy at these two specific sites.

In addition to global data for 2000-2002 and 2005-2007 from the World Data Centre for Precipitation Chemistry, we also evaluate our estimates of NOX deposition over the long-term (1996-2014) using continuous observations provided by regional networks. Figure 4 shows measured wet NO3 deposition over North America, Europe, and East Asia for locations where at least 15 years of quality-controlled annual data are available. These long-term mean observations demonstrate many of the same spatial patterns as the time slices from 2000-2002 and 2005-2007. In North America, a relatively smooth gradient is observed from low deposition in the west to high deposition at sites in the east. In Europe, the highest measured long-term mean wet NO3 deposition occurs at a coastal site in southern Norway, at a site just east of Copenhagen, and at locations in northern Italy and in Switzerland. At higher latitude sites, deposition is lower. Across the eastern Asia network, the measurements show highest deposition at sites in Southeast Asia (e.g. at a location between Kuala Lumpur and Singapore, and another in the vicinity of Jakarta) and in Japan. The lowest long-term mean deposition
occurs at high latitude sites along the border of Russia and Mongolia, while moderate to high deposition is measured on the coast of eastern China.

In general, our satellite-constrained estimate reflects the spatial variability that is seen in the measurements. Globally, the correlation between measured NO$_3^-$ deposition and our estimated wet NO$_3^-$ deposition is excellent ($r = 0.83$, NMB = -7.7%, N = 136 gridded model-data pairs). The vast majority of pairs (> 85%) agree to within 50% of each other. For the individual regions, normalized mean bias in our estimate is smallest over North America (NMB = +2.4%), and higher over Europe and East Asia (NMB = -32% and -25% respectively). The spatial correlation over each region is strong ($r = 0.89$, r = 0.87, and r = 0.69 for North America, Europe, and East Asia respectively), but sample sizes over Europe (N = 16) and East Asia (N = 11) are small so we emphasize caution in the interpretation of the statistics for these two regions. The lack of continuous measurement coverage even in parts of the world with routine network observations highlights the imperative of using other novel observational constraints on deposition (such as the global satellite observations of NO$_2$ used here).

### 3.2 Trends in Global NO$_y$ Deposition from 1996 to 2014

Our long-term satellite-constrained estimate of NO$_y$ deposition facilitates a unique and up-to-date investigation of the changes in NO$_y$ deposition around the world. We calculate linear trends in annual NO$_y$ deposition using the nonparametric Sen’s method (Sen, 1968), and test for significance with the nonparametric Mann-Kendall method (Kendall, 1975; Mann, 1945). We treat increasing or decreasing trends as significant if $p < 0.01$. Given that this is a test for linear trends, regions where shorter-term trends in deposition may have changed signs over the period of study could result in erroneous or insignificant trends. Below we discuss particular regions where this is the case.

Figure 5 shows the long-term annual and seasonal trends calculated from our satellite-constrained estimate of total NO$_y$ deposition across 1996-2014 (hatching indicates statistical significance). Figure 6 highlights timeseries of total NO$_y$ deposition over three specific regions covering parts of North America, Western Europe, and East Asia (as outlined in dashed boxes in the top panel of Fig. 5).

Substantial decreases are seen throughout North America extending over the Atlantic Ocean to remote regions. The timeseries for this region (Fig. 6, left) shows that NO$_y$
deposition decreased by almost 40% from 6.4 Tg N yr\(^{-1}\) in 1996-1998 to 3.9 Tg N yr\(^{-1}\) in 2012-2014. The steepest local decline in the world appears over the Ohio River Valley area, with a maximum near Pittsburgh where NO\(_x\) deposition decreased by -0.6 kg N ha\(^{-1}\) yr\(^{-2}\). NO\(_x\) deposition near Pittsburgh decreased from consistently exceeding 15 kg N ha\(^{-1}\) yr\(^{-1}\) during 1996-2000, to below 6 kg N ha\(^{-1}\) yr\(^{-1}\) by 2014. The strong decrease in the northeast is consistent with other long-term observational studies for the US (Sickles II and Shadwick, 2007, 2015). Studies of US NO\(_x\) emissions derived from satellite observations have also highlighted the remarkable success of emission controls (Duncan et al., 2013; Russell et al., 2012). Our constrained estimate has the steepest declines during the summer (Fig 5, JJA), restricted tightly to the source regions. This also agrees with long-term observations showing the strongest reductions in the summer (Sickles II and Shadwick 2015), consistent with the shorter lifetime of NO\(_x\) and efficient dry deposition of NO\(_x\) over the forested eastern US. We find significant decreases far downwind over the Atlantic Ocean during the other months, when NO\(_x\) can be transported farther. The steep change in NO\(_x\) deposition in the eastern US over the last 20 years may have important consequences on tree mortality rates in the region, which have been demonstrated to be very sensitive to NO\(_x\) deposition in the range of 5-15 kg N ha\(^{-1}\) yr\(^{-1}\) (Dietze and Moorcroft, 2011). The steeply decreasing trends across the US in our satellite-derived NO\(_x\) also support the increasing dominance of reduced nitrogen in total nitrogen deposition evidenced by observations (Li et al., 2016) and model predictions (Ellis et al., 2013).

We find a small but statistically significant positive trend in NO\(_x\) deposition (+0.06 kg N ha\(^{-1}\) yr\(^{-2}\)) in northern Alberta, Canada, dominated by the trend in JJA. The region is downwind of development in the Canadian oil sands, which has seen notable changes in NO\(_2\) column abundance as observed from space (McLinden et al., 2012). We estimate that deposition of NO\(_x\) in this area was at a maximum of 3.4 kg N ha\(^{-1}\) yr\(^{-1}\) in 2011 (up from 1.3 kg N ha\(^{-1}\) yr\(^{-1}\) in 1996-1997), and has since declined to 1.6 kg N ha\(^{-1}\) yr\(^{-1}\) by 2014. Elsewhere in Canada we estimate that NO\(_x\) deposition has decreased in the south and east parts of the country, consistent with observational analyses (Zbieranowski and Aherne, 2011).

Declines in NO\(_x\) deposition are also found across Europe, but statistical significance tends to be limited to western continental Europe and the United Kingdom (while changes in the south, north, and eastern countries tend to be insignificant). According to the timeseries for this region (Fig. 6, middle), NO\(_x\) deposition decreased by about 15% (from 2.5 Tg N yr\(^{-1}\)
in 1996-1998 to 2.1 Tg N yr$^{-1}$ in 2012-2014). We find the steepest local trends (-0.1 kg N ha$^{-2}$ yr$^{-1}$) in eastern Germany and southern UK, where NO$_y$ deposition in 2012-2014 decreased by 20% compared to 1996-1998. Previous satellite constraints on NO$_x$ emissions established that NO$_x$ emissions in France, Germany, Great Britain, and Poland have declined since 1996 while emissions in Greece, Italy, Spain, and the Ukraine for example have either stayed constant or increased (Konovalov et al., 2008). The local variability in emission trends leads to notable transboundary impacts. For example, our simulation predicts no net trend in NO$_y$ deposition over the Ukraine; but we find this is a result of opposing trends in dry (increasing) and wet (decreasing) deposition. This would be explained by increasing local emissions but decreasing transport from upwind. Similarly, we find significant increases in dry deposition in parts of western Russia but no significant trend in wet deposition.

Large increases in NO$_y$ deposition are found throughout Asia, concentrated especially in eastern China and parts of Southeast Asia. Figure 6 shows the timeseries of wet and dry NO$_y$ deposition within the rectangular region outlined in Figure 5 that encompasses eastern China and part of the adjacent ocean. We find that NO$_y$ deposition in the region increased by 65% from 5.2 Tg N yr$^{-1}$ in 1996-1998 to 8.6 Tg N yr$^{-1}$ in 2012-2014. The timeseries also shows that NO$_y$ deposition decreased after peaking around 9.3 Tg N yr$^{-1}$ in 2011-2012. We find that the steepest increasing local trends in the world appear in eastern China, and in the Pearl River Delta region (up to +0.6 kg N ha$^{-1}$ yr$^{-2}$). In fact, deposition in the Pearl River Delta region is the highest in the world for most of our record, exceeding 20 kg N ha$^{-1}$ yr$^{-1}$ every year from 2003-2014 (almost doubling from just over 11 kg N ha$^{-1}$ yr$^{-1}$ in 1996). The trends in deposition over China are largest in summer when the NO$_x$ lifetime is short, with more obvious indications of increasing NO$_y$ transport/export in the spring months (Fig 5, MAM).

The substantial increase in NO$_x$ emissions throughout East Asia has been inferred from satellite instruments in several previous studies (Mijling et al., 2013; Richter et al., 2005). The timing and extent of the reversal in NO$_y$ deposition that we see is also consistent with observed NO$_2$ columns over eastern China derived from OMI (de Foy et al., 2016; Krotkov et al., 2016). The ability of our satellite-constrained NO$_y$ deposition estimate to capture this sudden dramatic decrease over China, in contrast with previous projections (e.g. the RCP2.6, RCP4.5, and RCP8.5 projections to 2030, Lamaque et al. 2013), emphasizes an attribute of the satellite constraint.
Small statistically significant decreasing trends are found over the biomass-burning dominated source regions of Africa. The decrease in NO$_y$ deposition of about ~3% yr$^{-1}$ relative to the long-term mean in Northern Africa is consistent with the most recent GFED4 inventory from 1997 to 2014 (http://www.globalfiredata.org/), which has fire NO$_x$ emissions decreasing at a rate of about 3% yr$^{-1}$ in this region. In contrast, we also estimate a similar decrease in Southern Africa that is not represented in the recent GFED4 emission timeseries. A reduction in NO$_2$ column abundance in this region (observed by GOME and SCIAMACHY) was also reported by van der A. (2008). They postulate that this decline could be a result of deforestation leading to less biomass burning, but changing NO$_x$ emission factors from biomass burning could also potentially explain contribute to the trend.

Despite the large regional trends described above, we find that global deposition changed very little between 1996-1998 (56.1 Tg N yr$^{-1}$) and 2012-2014 (58.5 Tg N yr$^{-1}$) due to the opposing changes in different regions. Total NO$_y$ deposition was lowest in 2006 (50.5 Tg N), and peaked at 60.8 Tg N in 2012. Since then, it appears that global NO$_y$ deposition may be on the decline. Future observations in the coming years will be needed to establish whether this most recent decline is robust or temporary.

Figure 7 shows the calculated long-term trends in the measured wet NO$_3^-$ deposition for locations across North America, Europe, and East Asia where at least 15 years of quality-controlled annual data are available (the coverage of these observation is the same as in Figure 4). The observations over North America show the gradient in trend from negligible in the west to steeply and significantly decreasing in northeastern US and southeastern Canada. The steepest observed statistically significant trend (-0.18 kg N ha$^{-1}$ yr$^{-2}$) occurs east of Detroit in southwestern Ontario, Canada. In Europe, only one of the gridded observations has a statistically significant trend (-0.07 kg N ha$^{-1}$ yr$^{-2}$), located near the border of Denmark and Germany. The other locations in Europe show statistically insignificant trends over the long-term. In East Asia, we also find that most of the stations record statistically negligible trends over the long term (only two of the 11 gridded observations have significant trends). The steepest observed trend in this region (+0.39 kg N ha$^{-1}$ yr$^{-2}$) is found near Kuala Lampur and is statistically significant.

We compare the long-term trends in these measurements with our satellite-constrained trends in wet NO$_3^-$ deposition. We find a similar spatial gradient in North America, and the same magnitude of declines through the northeast US and southern Ontario (-0.12 to -0.16 kg
N ha\(^{-1}\) yr\(^{-2}\)). Over Europe, our estimates have low statistical significance in the trends throughout much of this region, consistent with the observations. Where we do see statistical significance (northern United Kingdom, southern Denmark, and in some central/eastern European countries), observations are not available over the long-term for evaluation. In East Asia, our satellite-constrained estimated trends show statistically significant increases throughout much of the region (in contrast to most of the available observations). The trend over Kuala Lumpur is significant and positive (+0.24 kg N ha\(^{-1}\) yr\(^{-2}\)) as expected from the available measurements.

We again emphasize the small sample size in Europe (N = 16) and East Asia (N = 11). Moreover, in many cases trends in one (or both) datasets are small and/or insignificant. For these reasons, we focus on comparing the confidence intervals of the measured and satellite-constrained trends. We find that for 129 of the 136 gridded pairs (> 90% of the data), the 95% confidence intervals overlap; of the pairs for which the intervals do not overlap, 3 (out of 109) occur in North America, 1 (out of 16) in Europe, and 3 (out of 11) in East Asia. For a large majority of the data in all three regions we therefore conclude that the satellite-derived trends are not significantly different from the trends inferred with ground-based measurements. Continued long-term measurements with better spatial coverage are imperative to better evaluate long-term estimates of global NO\(_y\) deposition especially throughout Europe and East Asia (but also in other parts of the world where long-term coverage is not available at all).

### 3.3 Changes in Continental Export of NO\(_y\)

NO\(_y\) deposition is a transboundary, and even intercontinental, issue (HTAP, 2010). In a multi-model study, Sanderson et al. (2008) found that between 3-10% of NO\(_x\) emissions from Europe, North America, South Asia, and East Asia are ultimately deposited over foreign regions. Long range transport events of NO\(_2\) alone can be systematically detected by satellite observations (Zien et al., 2014). Here we extend such studies using our satellite-constrained long-term estimates of annual NO\(_y\) deposition to evaluate how the amount of NO\(_x\) exported from specific regions (i.e. the net balance between emissions and deposition over a land area) has changed over the last two decades.

Decreases in NO\(_y\) export over the Atlantic Ocean from North America and increases in export over the western Pacific Ocean from East Asia are evident in Figure 5. We find that net export of NO\(_x\) from North America via atmospheric transport has decreased by more than
40% (from 2.5 Tg N yr\(^{-1}\) in 1996, to 1.4 Tg N yr\(^{-1}\) in 2014). In contrast, we find that export of NO\(_x\) from Asia increased by 40% from 3.3 Tg N in 1996, to a maximum of 4.7 Tg N in 2011, with a subsequent decrease to 3.8 Tg N by 2014. As a result of these opposing trends, total deposition to the global oceans has changed remarkably little (25.0 Tg N yr\(^{-1}\) in 1996-1998 compared to 24.4 Tg N yr\(^{-1}\) in 2012-2014), but has experienced substantial regional redistribution.

NO\(_y\) export from North America has received considerable attention. Urban plumes from the eastern US that are transported across the North Atlantic for several days could still contain 20-50 ppb of reactive nitrogen oxides (Neuman et al. 2006). A recent detailed GEOS-Chem study of nitrogen deposition over the US estimated that net annual export of NO\(_x\) was around 38% of NO\(_x\) emissions (or 2.5 Tg N) for 2006-2008 (Zhang et al. 2012). We estimate a similar fraction of export from the continental US using our observationally-constrained simulation (34% ± 2% from 1996-2014), with a small decreasing trend from 35% in 1996-1998 to 32% in 2012-2014. As a result of declining emissions we find that absolute export from the continental US decreased by 50% from 2.9 Tg N in 1996 to 1.5 Tg N in 2014. We find declines in NO\(_y\) deposition across the Atlantic Ocean, with small though statistically significant declines as far downwind as southern Greenland. The decreases downwind of the continent are clearest and most significant in the winter, spring, and fall (Fig. 5b, c, and e) while the trends are more local in the summer (when the NO\(_x\) lifetime is short and when midlatitude wind speeds are weaker.

We similarly calculate the net imbalance between NO\(_x\) emissions and NO\(_y\) deposition over western European countries and find a decrease of almost 40%, from 2.2 Tg N to 1.3 Tg N. We calculate mean export of NO\(_x\) emissions from western European countries to be 45% ± 4%, with a notable decreasing trend from 50% in 1996-1998 to 40% in 2012-2014. In contrast to the continental US, we also find a notable decrease in the fraction of emissions that are exported from western European countries (from 50% in 1996-1998 to 40% in 2012-2014). As a result, the decrease in net export is steeper than the decrease in emissions from the region. As alluded to in Section 3.2, the decrease in NO\(_x\) export from some western European countries has likely compensated for increases in emissions in some of the central/eastern European countries, and in western Russia, where we find dry deposition has significantly increased, but wet deposition has decreased or shows no significant net trend.
Reactive nitrogen transport from Asia has previously been shown to contribute to $O_3$ production across the mid-latitude Pacific reaching as far as the west coast of North America (Walker et al., 2010; Zhang et al., 2008), and major NO$_x$ transport events from China can be indirectly observed by NO$_2$ columns (Lee et al., 2014). Our satellite-constrained estimate predicts that export from China alone (24% ± 4% of emissions) more than tripled from 1.0 Tg N in 1996 to a maximum of 3.5 Tg N in 2011, then decreased to 2.5 Tg N by 2014. **We estimate that an average of 24% ± 4% emissions from China are exported, varying over time from as little as 15% of emissions in 1998 to a maximum of 31% of emissions in 2011 (with an overall increasing trend). Zhao et al. (2017) used a higher resolution (0.5° x 0.667°) GEOS-Chem simulation and estimated that 36% of China’s NOx emissions over 2008-2012 are exported.** We calculate an export fraction of around 27% for the same time period. The discrepancy between the two estimates may be attributed to the coarser horizontal resolution of our simulation (2° x 2.5°), pointing to important resolution-dependent effects in global simulations of deposition. Other factors may include the use of different NOx emissions (our satellite-constrained emissions indicate rapid change over this period of time), and the treatment of adjacent oceans.

Zhao et al. (2015) used GEOS-Chem to explore nitrogen deposition to the northwestern Pacific Ocean off the coast of China from 2008-2010. They estimated total (wet + dry) NO$_x$ deposition of 6.9 kg N ha$^{-1}$ yr$^{-1}$ and 3.1 kg N ha$^{-1}$ yr$^{-1}$ to the Yellow Sea and the South China Sea respectively. Our simulation predicts that NO$_x$ deposition to the same regions of the Yellow Sea increased from 5.1 kg N ha$^{-1}$ yr$^{-1}$ in 1996 to 9.5 kg N ha$^{-1}$ yr$^{-1}$ by 2012 and to the South China Sea from 2.8 kg N ha$^{-1}$ yr$^{-1}$ in 1996 to 4.3 kg N ha$^{-1}$ yr$^{-1}$ in 2011. Subsequent declines in the following years will hopefully have encouraging implications for nitrogen availability and the incidence of algal blooms in these regions (Hu et al., 2010).

Export of pollution from China has been shown to influence deposition over Japan in particular (e.g. Lin et al., 2008). Using observations of wet nitrate deposition, Morino et al. (2011) report increases throughout Japan from 1989-2008, and attribute this trend largely to transport from China. Likewise, integrated NO$_x$ deposition over Japan increased ($p < 0.01$) in our satellite-constrained estimate. In fact, we find that Japan transitioned from a net “exporter” of NO$_x$ over 1996-2006 (emissions exceeded local deposition by up to 24%) to a net “importer” of NO$_x$ over 2007-2014 (local deposition exceeded emissions by up to 20%). The increase in deposition was dominated by statistically significant increases in wet
deposition in some parts of the country. We find the increase over Japan is most uniform during the spring (Fig. 5, MAM), consistent with transport from China being pronounced during the spring season (Tanimoto et al., 2005). Nevertheless, the impacts of local NOx controls can also be important. Dry deposition dominates the decline in annual NOy deposition just west of Tokyo. Declines are seen throughout the southern part of the country during both the summer and fall seasons (Fig. 5, JJA and SON). These results demonstrate the indirect relationship between local emissions and local deposition of NOy for regions influenced by atmospheric transport, and also show how long-term trends can depend strongly on the season and process (wet or dry deposition).

3.4 Sensitivity of NOy Deposition to NH3 Emissions

The transport and ultimate deposition of oxidized nitrogen may be tightly coupled with the reduced nitrogen (NHx = NH3 + NH4+) and sulfate systems, due to the formation of NH4NO3 aerosol that becomes favorable once all H2SO4 has been neutralized (i.e., if there is “excess” NH3). Examples of the resulting non-linearity between PM2.5 concentrations and precursor emissions have been noted in the literature (Banzhaf et al., 2013; Derwent et al., 2009; Fowler et al., 2005). The formation of NH4NO3 aerosol at the expense of HNO3 with changing excess ammonia could therefore conceivably change the atmospheric lifetime of NOy at the surface; accumulation mode aerosol may have a dry deposition lifetime of days whereas HNO3 tends to have a dry deposition lifetime of shorter than a day. As a result, the predicted footprint of source impacts is sensitive to NH3 emissions (Lee et al., 2016).

Contemporary emissions of NH3 are highly uncertain (Reis et al., 2009), so we perform a sensitivity experiment by perturbing (increasing) all anthropogenic and natural NH3 emissions everywhere in the model by 25% for the year 2012. Predicted NOy deposition from this simulation is compared to the predicted NOy deposition in the 2012 simulation where NH3 emissions were not perturbed. Since we have not altered the emissions of oxidized nitrogen, simple mass balance dictates that increases in deposition over some regions will be countered by decreases elsewhere. Our perturbation is therefore to be interpreted as an experiment that tests how accurately the spatial pattern in NOy deposition at our model resolution can be predicted, given some uncertainty in NH3 emissions. Given the horizontal resolution of our simulation (2.5° x 2.0°), we acknowledge that our estimates of the sensitivity
of NO$_y$ deposition to perturbations in NH$_3$ emissions may underestimate the importance of those interactions at finer spatial scales.

Figure 8 shows the results of this experiment. The sensitivity of NO$_y$ deposition to an increase in NH$_3$ emissions is positive or negative depending on the region, while net deposition over the global domain does not change (to within 1-2%). Over the continents, the sensitivity in total (wet + dry) NO$_y$ deposition to the 25% perturbation in NH$_3$ emissions tends to be less than ±5%, with a few exceptions. We find differences in NO$_y$ deposition on the order of 10% over parts of high-latitude Russia, northwest and central Africa, eastern China, southern South America, and Australia. However, with the exception of China, these are also regions where deposition is relatively low. We conclude that for most regions of interest, our satellite-constrained estimates of NO$_y$ deposition over the continents and their trends will not be severely impacted by uncertainty in the NH$_3$ inventories.

Notably, the difference exceeds +50% over Myanmar, suggesting that simulated NO$_y$ deposition over this country is extremely sensitive to changes in NH$_3$ emissions. It is clear from Figure 8 that this results from a high sensitivity in dry deposition (middle panel) instead of wet deposition (bottom panel). Myanmar has some of the lowest estimated NH$_3$ emissions in all of South and East Asia (at least an order of magnitude lower than surrounding India, China, and Thailand), so this sensitivity reflects changes in the upwind emissions and subsequent transport of NO$_y$. We find the opposite sensitivity in nearby Cambodia, where the sensitivity of dry NO$_y$ deposition to a 25% perturbation is NH$_3$ emissions is -50%.

Over the oceans, the sensitivity of NO$_y$ deposition to the 25% increase in NH$_3$ emissions is generally low (< ±5%), with the expected exceptions in areas that are directly offshore from major continental source regions. In the North Atlantic Ocean east of Canada and Greenland, and in the North Pacific Ocean off the coasts of China, Japan and in the South China Sea, the sensitivity of NO$_y$ deposition is between 5-20%. Our predicted decrease in dry NO$_y$ deposition to the Yellow Sea given an increase in NH$_3$ emissions is consistent with previous adjoint analyses showing increased NO$_y$ dry deposition in this region with a decrease in Asian NH$_3$ emissions (Zhao et al. 2015). Likewise, the sensitivity of deposition to the Mediterranean Sea is between 10-20%. The differences in NO$_y$ deposition over the oceans results from sensitivity in both dry and wet deposition (although in the case of the Mediterranean it is dominated by dry deposition). We conclude that although changes (or uncertainties) in NH$_3$ emissions can impact the distance of transport and deposition to oceans
downwind of the major NO\textsubscript{x} sources, the absolute magnitude of deposition is low where the
sensitivity of NO\textsubscript{y} deposition to NH\textsubscript{3} is relatively high.

### 3.5 Other Considerations

A number of other uncertainties are important in an inversion of satellite NO\textsubscript{2} columns
to calculate surface NO\textsubscript{x} emissions and simulate of long-term NO\textsubscript{y} deposition. These can
depend on, for example, the choice of inversion approach, errors in the satellite retrieval, and
uncertainties in model processes (e.g. emissions, boundary layer mixing, chemical NO\textsubscript{3} sinks,
meteorology, and dry deposition).

Cooper et al. (2017) found that the finite mass balance inversion approach used here can
be improved upon by using an iterative method that performs with similar accuracy as a four-
dimensional variational data assimilation. Multi-constituent data assimilation also shows
considerable promise for constraints on surface NO\textsubscript{3} emissions (Myazaki et al. 2017). Satellite
retrieval algorithms continue to develop with advances that will improve the accuracy of
future estimates of satellite-constrained NO\textsubscript{y} deposition.

Uncertainties in model processes are also of interest. For example, uncertainties in the
chemical sink of NO\textsubscript{3} alone (e.g. the rate of HNO\textsubscript{3} formation, heterogeneous loss of N\textsubscript{2}O\textsubscript{5}
on to aerosol) can have a substantial impact on top-down emissions estimates (Stavrakou et
al., 2013), suggesting more fundamental work in constraining these processes is required. Lin
et al. (2010) found that top-down NO\textsubscript{3} emissions estimates over East Asia are sensitive to
other model uncertainties including planetary boundary layer mixing scheme, lightning
emissions, diurnal profile of emissions, and a-priori NO\textsubscript{x}, CO, and VOC emissions.
Uncertainties in model meteorology are also important. For example, the MERRA
precipitation fields used in our study are known to correlate weakly with observational
datasets (Rienecker et al., 2011), but improvements can be expected from MERRA-2 due to
the inclusion of gauge- and satellite-based precipitation corrections (Reichle et al., 2017).
Finally, dry deposition schemes are also highly variable among models (Flechard et al., 2011;
Hardacre et al., 2015), and future work in dry deposition evaluation should be a priority.

Nonetheless, despite these uncertainties we find a high degree of consistency between
observations and our predictions in the long-term changes to deposition. Evidence continues
to emerge about potential biases in bottom-up inventories (e.g. Travis et al. 2016), and our
observational constraint on NO\textsubscript{x} emissions mitigates against such biases. We expect
4 Conclusion

\(\text{NO}_y\) deposition represents about half of the total reactive nitrogen deposited to Earth’s surface. Even in the US where nitrogen oxide emissions have decreased substantially, constituents of \(\text{NO}_y\) remain major contributors to the nitrogen deposited in areas of concern (Lee et al. 2016; Li et al. 2016). We applied \(\text{NO}_2\) observations from multiple satellites over 1996-2014 together with the GEOS-Chem chemical transport model to estimate long-term changes to reactive nitrogen oxide deposition around the world. Given the effective global mass balance between \(\text{NO}_x\) emissions and deposition of reactive nitrogen oxides, we show that satellite constraints on \(\text{NO}_x\) emissions can provide a powerful top-down constraint on deposition in order to evaluate long-term changes worldwide. Observations from the GOME, SCIAMACHY, and GOME-2 satellite instruments have provided continuous global coverage over the last 20 years, allowing observational constraints on \(\text{NO}_y\) deposition that enhance the poor spatial coverage of ground-based deposition measurements.

We find substantial variability in regional trends of \(\text{NO}_y\) deposition. \(\text{NO}_y\) deposition declined most steeply throughout the northeastern United States at a rate of up to -0.6 kg N ha\(^{-1}\) yr\(^{-2}\), but has also decreased significantly throughout most of the country and in southern Canada. In Europe, statistically significant declines at a rate of up to -0.1 kg N ha\(^{-1}\) yr\(^{-2}\) are seen over some western countries. On the other hand, \(\text{NO}_y\) deposition has increased substantially throughout East Asia, exceeding a rate of +0.6 kg N ha\(^{-1}\) yr\(^{-2}\) in some parts.

Since reductions in deposition over some regions were counteracted by increases in others, global \(\text{NO}_y\) deposition did not change considerably over the long term. However, we find that global \(\text{NO}_y\) deposition could now be on the decline overall, since deposition in Asia peaked around 2010-2012. The ability to resolve the striking recent decline in \(\text{NO}_y\) deposition in China (despite prior projections of increasing \(\text{NO}_x\) emissions) demonstrates one of the attributes of using a satellite-based constraint. Future observations will be important in evaluating whether this trend persists.

We find that changes over the last two decades in the export of reactive nitrogen oxides via atmospheric transport have impacted countries downwind of source regions. Export from
North America has decreased by at least 40%, while export from Asia has increased by the same relative amount. We find evidence that decreases in NO\textsubscript{x} export from some western European countries have counteracted increases in local emissions from some eastern/central European countries, resulting in negligible net change in NO\textsubscript{y} deposition over the long term. Likewise, Japan is highly sensitive to changes in export from China, but this depends strongly on the season and whether wet and dry deposition are both considered. While uncertainty in NH\textsubscript{3} emissions can impact the footprint of NO\textsubscript{y} export and deposition, we show that this sensitivity is small in most regions of concern.

Direct measurements of deposition are sparse, inhibiting evaluation. This is especially challenging for global simulations, where individual measurements may not necessarily be regionally representative. Nevertheless, we find that for the vast majority of locations our satellite-derived trends are largely consistent with the observed trends. Expanded coverage of ground-based observations over the long-term is needed to more comprehensively evaluate long-term estimates of global NO\textsubscript{y} deposition. This need also motivates the value of using alternative observational constraints like the satellite NO\textsubscript{2} columns as presented here.

Forthcoming satellite observations of NO\textsubscript{2} at higher spatial resolution (e.g. TROPOMI (Veefkind et al., 2012)) and with diurnally varying observations (e.g. TEMPO (Zoogman et al., 2017), Sentinel-4, and GEMS) will offer increasingly robust constrains on NO\textsubscript{x} emissions that affect NO\textsubscript{y} deposition. Satellite observations of NH\textsubscript{3} (e.g. Van Damme et al., 2014) may offer additional opportunities to constrain the reactive nitrogen budget. Higher resolution global modeling will also be an important development to accurately account for non-linear NO\textsubscript{2} losses in global emission inversions (Valin et al., 2011).

Our satellite-constrained estimates of NO\textsubscript{y} document interannual changes over the past two decades worldwide. We expect that this information will be useful in future research into the impacts of nitrogen deposition to important biodiversity hotspots, in regions dealing with excessive nitrogen inputs leading to algal blooms, or estimating the changing impacts of nitrogen deposition on global carbon uptake.

Appendix 1

We simulate atmospheric chemistry from 1996 to 2014 using the GEOS-Chem chemical transport model (www.geos-chem.org) v9-02. Our simulations are driven with the
MERRA meteorological product at a global horizontal resolution of 2.5° x 2.0° and 47 vertical layers. GEOS-Chem includes detailed HOx-NOx-VOC-O3-aerosol chemistry (Bey et al. 2001; Park et al. 2004), with isoprene chemistry following Paulot et al. (2009a, 2009b) and gas-aerosol partitioning for the sulfate-nitrate-ammonium system calculated according to the ISORROPIA II equilibrium model (Fountoukis and Nenes, 2007). Gas-aerosol phase coupling occurs via N2O5 uptake (Evans and Jacob, 2005) and HO2 uptake (Mao et al., 2013) in addition to other heterogeneous chemistry (Jacob, 2000) and aerosol effects on photolysis frequencies (Martin et al., 2003). Our simulations use timesteps of 15 minutes for transport and convection, and 30 minutes for emissions and chemistry.

Removal by wet deposition occurs through scavenging in moist convective updrafts, as well as in-cloud and below-cloud scavenging during large-scale precipitation for water-soluble aerosol and gases (Amos et al., 2012; Liu et al., 2001). Removal by dry deposition is calculated based on the widely used resistance-in-series formulation from Wesely (1989), described for GEOS-Chem in Wang et al. (1998) and Zhang et al. (2001) for aerosol. Dry deposition of NOx over the United States was recently explored and evaluated in detail by Zhang et al. (2012).

Anthropogenic emissions are prescribed by the NEI 2005 inventory for the United States (http://www.epa.gov/ttnchie1/trends/), the CAC inventory for Canada (http://www.ec.gc.ca/pdb/cac/), the BRAVO inventory for Mexico (Kuhns et al., 2005), the EMEP inventory for Europe (http://www.emep.int/), and Zhang et al. (2009) for China and Southeast Asia. Elsewhere, the EDGAR v3 emission inventory is used for anthropogenic NOx, CO, and SOx (Olivier et al., 2005), the GEIA inventory for NH3 (Bouwman et al., 1997), and the RETRO inventory for VOCs (Hu et al., 2015). Monthly scaling of NOx emissions is included in North America (based on the VISTAS inventory), Europe (based on the EMEP inventory), and Asia (based on the Zhang et al. (2009) inventory). Monthly scaling of EDGAR emissions is based on the seasonality from the Global Emission Inventory Activity (Benkovitz et al 1996). Aircraft emissions are from the AEIC inventory (Stettler et al., 2011).

Scale factors based on energy statistics following van Donkelaar et al. (2008) are used to scale NOx, CO and SOx emissions between 1996 and 2010 for years when the emissions are unavailable from the inventory. For other species and for emissions beyond 2010, the closest available year is used. Biogenic VOC emissions are calculated using the MEGAN model (Guenther et al., 2006). Biomass burning emissions are according to the GFED3 inventory.
(Mu et al., 2011). Soil NO$_x$ is calculated using the Berkeley-Dalhousie Parameterization (Hudman et al., 2012). Lightning NOx is implemented according to Murray et al. (2012). These a-priori surface NO$_x$ emissions are overwritten by our satellite-derived top-down estimates in the assessment of NO$_y$ deposition.

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Table 1: Global top-down NOx emissions calculated using the finite mass balance inversion approach with observations from GOME, SCIAMACHY, and GOME-2.

<table>
<thead>
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<th>Year</th>
<th>Global NOx Emissions (Tg N yr⁻¹)ᵃ</th>
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<tr>
<td>2014</td>
<td>54.0</td>
</tr>
<tr>
<td>Mean</td>
<td>55.6</td>
</tr>
</tbody>
</table>

ᵃ Includes anthropogenic HNO₃ flux of 2.3 ± 0.1 Tg N yr⁻¹.
Figure 1: Long-term (1996-2014) mean NO$_y$ deposition derived from the GEOS-Chem simulation constrained by satellite observations of NO$_2$ columns from the GOME, SCIAMACHY, and GOME-2 instruments (top). Mean ratio of simulated dry NO$_y$ deposition to total NO$_y$ deposition (bottom).
Figure 2: Annual wet NO$_3^-$ deposition from measurements available through the World Data Centre for Precipitation Chemistry, and from the GEOS-Chem simulation constrained with satellite observations of NO$_2$. Two time periods are represented: 2000-2002 and 2005-2007.
Figure 3: Scatter plot of the satellite-constrained simulated wet NO$_3^-$ deposition vs. measurements available through the World Data Centre for Precipitation Chemistry for specific subsets of the data. The red lines show the result of a reduced major axis linear regression. In the right column, the blue line shows the fit across all data and the red line shows the fit excluding the two circled data points that are discussed in the text (reported statistics refer to the red line fit).
Figure 4: Long-term (1996-2014) wet NO$_3^-$ deposition from available regional network measurements (top: NADP and CAPMON; middle: EMEP; bottom: EANet), and from the GEOS-Chem simulation constrained with satellite observations of NO$_2$. 
Figure 5. Long-term trend (1996-2014) in the satellite-constrained simulation of NO$_y$ deposition. (A) Annual mean; (B) December-January-February; (C) March-April-May; (D) June-July-August; (E) September-October-November. Diagonal hatching represents trend significance (p < 0.01). Hatching from top-left to bottom-right indicates a decreasing trend; hatching from bottom-left to top-right indicates an increasing trend.
Figure 6: Timeseries of annually integrated dry and wet NOy deposition over specific regions (North America, Europe, and East Asia) as defined by the dashed rectangles in Figure 5.
Figure 7: Long-term (1996-2014) trends in wet NO$_3^-$ deposition from available regional network measurements (as in Figure 4), and from the GEOS-Chem simulation constrained by satellite observations of NO$_2$. Closed circles around the measurements indicate significant trends (p < 0.01); hatching indicates statistical significance in the simulation.
Figure 8: Sensitivity of simulated NO$_y$ deposition in 2012 to a 25% perturbation (increase) in ammonia emissions in all grid boxes (shown separately for total deposition, dry deposition, and wet deposition).