

1 **Stratosphere-troposphere separation of nitrogen dioxide**
2 **columns from the TEMPO geostationary satellite**
3 **instrument**

4
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14
15 **Abstract**

16 Separating the stratospheric and tropospheric contributions in satellite retrievals of
17 atmospheric NO₂ column abundance is a crucial step in the interpretation and application of
18 the satellite observations. A variety of stratosphere-troposphere separation algorithms have
19 been developed for sun-synchronous instruments in low Earth orbit (LEO) that benefit from
20 global coverage, including broad clean regions with negligible tropospheric NO₂ compared to
21 stratospheric NO₂. These global sun-synchronous algorithms need to be evaluated and refined
22 for forthcoming geostationary instruments focused on continental regions, which lack this
23 global context and require hourly estimates of the stratospheric column. Here we develop and
24 assess a spatial filtering algorithm for the upcoming TEMPO geostationary instrument that
25 will target North America. Developments include using independent satellite observations to
26 identify likely locations of tropospheric enhancements, using independent LEO observations
27 for spatial context, consideration of diurnally-varying partial fields of regard, and a filter
28 based on stratospheric to tropospheric air mass factor ratios. We test the algorithm with LEO

1 observations from the OMI instrument with an afternoon overpass, and from the GOME-2
2 instrument with a morning overpass.

3 We compare our TEMPO field of regard algorithm against an identical global algorithm
4 to investigate the penalty resulting from the limited spatial coverage in geostationary orbit,
5 and find excellent agreement in the estimated mean daily tropospheric NO₂ column densities
6 ($R^2 = 0.999$, slope = 1.009 for July and $R^2 = 0.998$, slope = 0.999 for January). The algorithm
7 performs well even when only small parts of the continent are observed by TEMPO. The
8 algorithm is challenged the most by east coast morning retrievals in the wintertime (e.g. $R^2 =$
9 0.995 , slope = 1.038 at 1400 UTC). We find independent global LEO (corrected for time of
10 day) provide important context near the field-of-regard edges. We also test the performance of
11 the TEMPO algorithm without these supporting global observations. Most of the continent is
12 unaffected ($R^2 = 0.924$ and slope = 0.973 for July and $R^2 = 0.996$ and slope = 1.008 for
13 January), with 90% of the pixels having differences of less than $\pm 0.2 \times 10^{15}$ molecules cm⁻²
14 between the TEMPO tropospheric NO₂ column density and the global algorithm. For near-
15 real-time retrieval, even a climatological estimate of the stratospheric NO₂ surrounding the
16 field of regard would improve this agreement. In general, the additional penalty of a limited
17 field of regard from TEMPO introduces no more error than normally expected in most global
18 stratosphere-troposphere separation algorithms. Overall, we conclude that hourly near-real-
19 time stratosphere-troposphere separation for the retrieval of NO₂ tropospheric column
20 densities by the TEMPO geostationary instrument is both feasible and robust, regardless of
21 the diurnally-varying limited field of regard.

22

23 **1 Introduction**

24 Nitrogen dioxide (NO₂) and nitrogen oxides in general are central to atmospheric
25 chemistry in both the troposphere and stratosphere (Finlayson-Pitts and Pitts, 1999; Seinfeld
26 and Pandis, 2016). In the stratosphere, nitrogen oxides are a key player in ozone (O₃)
27 depletion chemistry. In the troposphere, photolysis of NO₂ is responsible for the production of
28 O₃ whose buildup is associated with negative human health, ecosystem, and radiative forcing
29 impacts. Emissions of nitrogen oxides are also linked to the production of secondary
30 inorganic aerosol with impacts on both health and global climate. Observations of NO₂ in the
31 atmosphere are therefore critical given its roles in air quality and atmospheric chemistry.

1 Satellite remote sensing of NO₂ from instruments in low Earth orbit (LEO) has offered
2 extraordinary insight into global nitrogen oxide processes. Among many applications,
3 observations from GOME (1996-2003), SCIAMACHY (2002-2011), OMI (2004-), and
4 GOME-2 (2007-) have contributed to understanding global and regional patterns in nitrogen
5 oxide emissions (e.g. Beirle et al., 2003; Duncan et al., 2013; Jaegle et al., 2005; Kononov
6 et al., 2008; Lamsal et al., 2011; Martin et al., 2003; Miyazaki et al., 2016; Richter et al.,
7 2005; Russell et al., 2012), evaluating ground-level air quality in the absence of traditional
8 monitoring data (e.g. Bechle et al., 2013; Boersma et al., 2009; Geddes et al., 2016; Lamsal et
9 al., 2008; McLinden et al., 2012), and constraining nitrogen oxide deposition out of the
10 atmosphere (e.g. Geddes and Martin, 2017; Jia et al., 2016; Nowlan et al., 2014). A key step
11 in these applications is the separation of stratospheric and tropospheric NO₂ from the total
12 column derived from the satellite observation, a process that can introduce substantial
13 uncertainty the final tropospheric column estimates (Beirle et al., 2016; Boersma et al., 2004;
14 Bucsela et al., 2013; Martin et al., 2002).

15 Separating the stratospheric and tropospheric contributions to the total column has been
16 performed using a number of approaches, varying in complexity and in the assumptions that
17 are made. The simplest approach is the Pacific reference sector method (Beirle et al., 2003;
18 Martin et al., 2002; Richter and Burrows, 2002) in which stratospheric NO₂ is treated as
19 longitudinally homogeneous so that stratospheric NO₂ in any location can be estimated by
20 using the measured NO₂ over the remote Pacific at the same latitude. Tropospheric NO₂ in the
21 reference sector might either be ignored altogether (e.g. Richter and Burrows, 2002) or
22 accounted for using a model estimate (e.g. Martin et al., 2002). While the treatment of zonal
23 invariance is reasonable for low- to mid-latitudes, stratospheric dynamics (especially in the
24 vicinity of polar vortices) raise concerns at higher latitudes of relevance for planned
25 geostationary missions.

26 Image processing and spatial filtering techniques are an extension of the reference sector
27 method (Bucsela et al., 2006, 2013; Leue et al., 2001; Valks et al., 2011; Velders et al., 2001;
28 Wenig et al., 2004), whereby stratospheric NO₂ is estimated by interpolating between regions
29 that are classified as having negligible tropospheric NO₂. This might be accomplished for
30 example by using only cloudy scenes over the oceans (e.g. Leue et al., 2001), or by applying a
31 pollution “mask” given prior estimates of tropospheric NO₂ (e.g. Bucsela et al., 2006; Valks
32 et al., 2011). Bucsela et al. (2013) proposed a masking scheme that combines a prior estimate

1 of tropospheric NO₂ with radiative transfer calculations to allow polluted pixels to remain if
2 the scene is cloudy (obscuring lower tropospheric NO₂), and exclude unpolluted regions
3 where tropospheric NO₂ signal may still be significant due to high tropospheric air mass
4 factors. An elegant variation of this spatial filtering approach is the STRatospheric Estimation
5 Algorithm from Mainz (STREAM), developed by Beirle et al. (2016). Instead of binary
6 masks based on arbitrary thresholds, STREAM applies a weighted convolution scheme where
7 cloudy observations are given a high weight and polluted observations (based on a prior
8 estimate) are given low weight. These spatial filtering approaches developed exclusively for
9 global observational coverage from LEO offer valuable guidance on the development of
10 geostationary stratosphere-troposphere separation algorithms.

11 Nadir observations are also used in assimilation approaches where model predictions of
12 the stratospheric NO₂ column density are adjusted towards the observed column density. For
13 example, stratosphere-troposphere separation in the Dutch NO₂ algorithm is achieved by
14 assimilating observed NO₂ columns with model NO₂ column predictions from the TM4
15 chemical transport model forced by ECMWF meteorological data (Boersma et al., 2007;
16 Dirksen et al., 2011). In that approach, modeled NO₂ profiles are convolved into line-of-sight
17 (“slant”) columns using averaging kernels, and the difference between modeled and observed
18 slant column densities are used to force the modeled columns to an “analysed” state. Using
19 the most recent observations available, the “analysed” state can be used in a forecast model
20 run to predict the stratospheric field for near-real time retrievals (Boersma et al., 2007).

21 In some cases, independent stratospheric observations may be used in the separation of
22 stratospheric and tropospheric NO₂. For example, the SCIAMACHY instrument made almost
23 coincident nadir and limb measurements (Bovensmann et al., 1999) and this matching was
24 exploited in algorithms by Beirle et al. (2010) and Hilboll et al. (2013). Even non-coincident
25 limb-nadir matching has been exploited for stratosphere-troposphere separation, as in the case
26 of OSIRIS and OMI (Adams et al., 2016). Sussmann et al. (2005) demonstrate how
27 simultaneous ground-based measurements (especially at mountain sites) could be applied for
28 stratosphere-troposphere separation algorithm validation.

29 To date, all of the above approaches to stratosphere-troposphere separation have been
30 developed using the large coverage of observations provided by instruments in LEO.
31 Questions remain about how well the separation can be performed without the global context
32 and where clean tropospheric background signals are limited. Stratosphere-troposphere

1 separation algorithms need to be evaluated and refined for the restricted field of regard of
2 future geostationary instruments such as TEMPO (Zoogman et al., 2017), Sentinel-4
3 (Veihelmann et al., 2015), and GEMS (Lasnik et al., 2014).

4 TEMPO (“Tropospheric Emissions: Monitoring of Pollution”), launching between 2019-
5 2021, will provide space-based measurements in geostationary orbit with a field of regard
6 over North America from southern Canada to Mexico City and the Bahamas (Zoogman et al.,
7 2017). The spectrometer has spectral ranges of 290-490 nm (at 0.57 nm resolution) and 540-
8 740 nm (at 0.2 nm resolution), allowing retrieval of tropospheric composition with fine spatial
9 resolution (up to 2.1 km North-South x 4.4 km East-West instantaneous field of view).
10 Scanning occurs from east to west, with hourly revisits. Among its standard products
11 available at roughly 4 km x 8 km spatial resolution will be hourly NO₂ column abundance.
12 Here, we develop a standard stratosphere-troposphere separation algorithm for the
13 observations of NO₂ from TEMPO, and examine in detail the potential information penalty
14 associated with the limited TEMPO field of regard compared to an identical global algorithm.

15

16 **2 Satellite Observations**

17 To develop and test our algorithm, we use data from two LEO instruments, with
18 afternoon and morning overpasses. We use NO₂ column densities derived from OMI on board
19 the Aura satellite launched in 2004. OMI is a nadir-viewing spectrometer in LEO crossing the
20 equator around 13:30 local time, with a variable horizontal resolution of 13 km x 24 km at
21 nadir. Line-of-sight (“slant”) columns are retrieved from spectral fitting of back-scattered and
22 reflected solar radiation within the 405-465 nm wavelength range, and corrected for
23 instrumental artifacts (Bucsela et al., 2013). We use the Version 3.0 Standard Product NO₂
24 retrieval (SPv3) from NASA (Krotkov et al., 2017, publicly available at
25 http://disc.sci.gsfc.nasa.gov/Aura/data-holdings/OMI/omno2_v003.shtml), including
26 stratospheric and tropospheric air mass factors provided with the data to relate slant and
27 vertical columns (Bucsela et al., 2013). We use the artifact-corrected slant column densities
28 (“destriping”) and the tropospheric and stratospheric air mass factors calculated for each
29 pixel. All data are first gridded to a 0.1° x 0.1° regular grid.

30 We also make use of NO₂ column densities derived from GOME-2, on board the MetOp-
31 A satellite launched in 2006. GOME-2 is another nadir-viewing spectrometer in LEO,
32 crossing the equator around 09:30 local time with a constant horizontal resolution of 80 km x

1 40 km in its default swath. Spectral fitting is performed within the 420-450 nm wavelength
2 range. Here we use the TM4NO2A retrieval (Boersma et al., 2004) version 2.3 data product
3 from KNMI (available from <http://www.temis.nl/airpollution/no2.html>) along with the
4 included air mass factors.

5 We restrict all data to solar zenith angles smaller than 80° to avoid exceedingly long path
6 lengths.

7

8 **3 Estimating Stratospheric NO₂ over the TEMPO Field of Regard**

9 Here we describe our approach to estimate the stratospheric NO₂ column in TEMPO
10 observations. As a foundation for our method, we begin with the approach used in the current
11 operational algorithm for OMI (Bucsela et al., 2013). This algorithm has demonstrated high
12 quality performance against validation data sets (Ialongo et al., 2016; Lamsal et al., 2014;
13 Bucsela et al., 2013), is computationally fast, and is suitable for near-real-time retrievals. Our
14 own implementation of this algorithm reproduces the operational global stratospheric NO₂
15 product well ($r = 0.99$ and a slope of 1.01). As described below, we build on this algorithm for
16 TEMPO by modifying certain smoothing/filtering steps, using a satellite-derived prior
17 estimate of tropospheric NO₂, incorporating observations surrounding the TEMPO field of
18 regard from independent LEO instruments, and by considering partial fields of regard relevant
19 to TEMPO.

20 Figure 1 shows the stepwise implementation of our TEMPO stratosphere-troposphere
21 separation algorithm for an example day in July. As a surrogate for TEMPO observations, we
22 begin by restricting the OMI total slant NO₂ column observations to the anticipated TEMPO
23 field of regard below a solar zenith angle threshold of 80° (Figure 1a). The expected coverage
24 of TEMPO extends from as far south as Mexico City, northward to include southern Canada
25 (covering as far north as the oil sands region in Alberta for example). The pattern along the
26 orbit tracks in Figure 1a results from the changing OMI viewing zenith angle (with higher
27 slant columns for larger viewing angles). Although we begin our implementation with the
28 OMI observations gridded to $0.1^\circ \times 0.1^\circ$, the TEMPO algorithm would be performed on the
29 individual TEMPO pixels. In other words, here we are treating our gridded OMI observations
30 as TEMPO pixels.

31 An initial estimate of the stratospheric vertical NO₂ column (V_{init}) can be obtained by:

$$1 \quad V_{init} = \frac{(S - S_{trop,prior})}{A_{strat}} \quad \text{Equation 1}$$

2 where S is the total slant column density, A_{strat} is the stratospheric air mass factor, and $S_{trop,prior}$
3 accounts for small contributions from the troposphere (Bucsela et al., 2013). Bucsela et al.
4 (2013) estimated the tropospheric contribution using model values. To provide a more
5 accurate constraint on tropospheric contributions, we use the monthly mean tropospheric NO₂
6 columns derived from independent GOME-2 observations as an initial a-priori tropospheric
7 NO₂ estimate. The GOME-2 observations were filtered using recommended quality flags and
8 retaining pixels with cloud radiance fraction less than 0.5, then gridded to the same resolution
9 as our OMI grid. This concept enables the use of spatial information observed from satellite,
10 and could be readily adapted to use TROPOMI observations at finer resolution. Ideally, an
11 independent LEO tropospheric estimate for as close to the TEMPO observation time would be
12 used. Nonetheless, diurnal variability in tropospheric NO₂ columns outside of source regions
13 tends to be small (Boersma et al., 2008), and in our case source regions are masked out in a
14 later step. The use of a satellite-derived a-priori reduces the use of chemical transport model
15 information in the stratosphere-troposphere separation algorithm (although we revert to a
16 model estimate if quality controlled satellite coverage is not available, e.g. due to
17 systematically high cloud fractions). We transform this satellite-derived a priori tropospheric
18 NO₂ vertical column ($V_{trop,prior}$) into slant column space using the tropospheric air mass factors
19 (A_{trop}) provided with the OMI data:

$$20 \quad S_{trop,prior} = V_{trop,prior} \cdot A_{trop} \quad \text{Equation 2.}$$

21 Figure 1b shows our initial estimate of stratospheric vertical NO₂ columns over the TEMPO
22 domain resulting from the combination of Equation 1 and 2. We already see that this
23 stratospheric NO₂ estimate varies predominately as a function of latitude, although
24 anomalously low values are seen over some urban centers (e.g. around Los Angeles, Chicago,
25 and New York) where the a-priori tropospheric NO₂ slant column is large.

26 To exclude locations where this initial stratospheric vertical column estimate is likely
27 biased, we make use of the masking approach from Bucsela et al. (2013). This is based on
28 eliminating pixels where tropospheric contamination is high (or where the initial stratospheric
29 vertical column estimate would exceed the actual stratospheric vertical column by some
30 reasonable value) by requiring:

1
$$\frac{S_{trop,prior}}{A_{strat}} < 0.3 \times 10^{15} \text{ cm}^{-2}$$
 Equation 3.

2 On a typical day in July, this means that contamination from the troposphere would be less
3 than ~10% percent of the stratospheric NO₂ estimate (which generally ranges from 2-4 x 10¹⁵
4 cm⁻² over the TEMPO field of regard). Figure 1c shows the result of this masking step. The
5 threshold removes all the urban regions with anomalously low values in Figure 1b, in addition
6 to many other areas. Sensitivity tests show that the final stratospheric NO₂ estimate varies by
7 less than 5% for changes in this threshold between 0.2 x 10¹⁵ or 0.4 x 10¹⁵ cm⁻², consistent
8 with the generally small sensitivity found by Bucsele et al. (2013). On this example day (and
9 for the month of July on average) the masking threshold of 0.3 x 10¹⁵ cm⁻² removes 55% of
10 the original data within the TEMPO field of regard. We find coverage is best over Canada and
11 over the Pacific Ocean, with less coverage over the rest of the continent and the Atlantic
12 Ocean. The original global algorithm removes ~28% of the available global data on average
13 for days in July, since tropospheric NO₂ columns are generally lower elsewhere in the world.

14 Since $S_{trop,prior}$ is calculated based on radiative transfer calculations (A_{trop}) in addition to
15 the a priori tropospheric NO₂ vertical column (Equation 2), this masking approach in principle
16 allows for polluted pixels to remain if the lower tropospheric signal is sufficiently suppressed
17 by clouds resulting in a low tropospheric air mass factor (or conversely excludes pixels with a
18 considerable tropospheric signal due to high surface reflectivity). We investigated the use of
19 explicitly cloudy scenes (cloud radiance fraction > 0.9), which could suppress the signal from
20 below. Mid-level clouds (600-400 hPa) are the least likely to contain significant NO_x mixed
21 in from the surface, or lightning NO_x associated with higher clouds. We find that most
22 (>75%) of the pixels that meet these criteria are already retained by our original masking
23 algorithm. Incorporating the remaining cloudy pixels to the masked data increases data
24 coverage by less than 1%. Given the uncertainties in retrieving cloud properties, uncertainties
25 in cloudy air mass factors, and the minimal added value of this dataset, we disregard adding
26 the remaining cloudy pixels to our algorithm.

27 In Bucsele et al. (2013), the remaining unmasked data are binned and un-filled bins are
28 interpolated using 2-dimensional averaging with a 30° longitude x 20° latitude moving
29 window. In our case, this step necessarily precludes information from outside the TEMPO
30 field of regard over the mostly pristine oceans from being used in the 2-D averaging. As we
31 will show, this leads to biases near the field of regard edges when compared to a global

1 algorithm, since the averaging window is disproportionately impacted by observations with
 2 continental influence. We reduce this bias by incorporating independent global observations
 3 from LEO that can provide context outside of the TEMPO field of regard. This approach
 4 exploits the independent LEO observations that are expected throughout the lifespan of
 5 TEMPO (e.g. GOME-2, TROPOMI).

6 Here, we employ GOME-2 observations as an independent dataset to estimate
 7 stratospheric NO₂ at GOME-2 overpass time outside the TEMPO field of regard by using an
 8 identical algorithm on this global data. We empirically transform the GOME-2 stratospheric
 9 NO₂ estimate to the TEMPO observation time (here, the OMI overpass time), using the
 10 climatological 30-day running mean local ratio of GOME-2 to OMI stratospheric NO₂. A
 11 similar observational or model climatology could readily be constructed with TEMPO data
 12 after launch based on the available LEO observations at the time. Figure 1d shows the
 13 outcome of this approach. The GOME-2 observations outside of the TEMPO field of regard
 14 retain the same magnitude and latitudinal gradient as the available observations within the
 15 TEMPO field of regard, suggesting that the additional context from an independent LEO
 16 instrument can be useful even when they are from a different time of day.

17 Before interpolating the unfilled bins, we apply a boxcar filter using a moving 15° x 10°
 18 window as follows. First, our boxcar filter returns a smoothed array using the following
 19 algorithm:

$$R_i = \frac{1}{w} \sum_{j=0}^{w-1} A_{i+j-w/2} \text{ where } \frac{(w-1)}{2} \leq i \leq N - \frac{(w+1)}{2}$$

Equation 4

21 where w is the smoothing width (in our case, defined in two dimensions by both a length and
 22 width), R_i is the i -th point in the smoothed data, and A_i is the i -th point in the original data. For
 23 data points where the neighborhood includes points outside the array, the nearest edge points
 24 are used to compute the smoothed result. The variance of the original data is also calculated
 25 using a similar algorithm. Any value that lies outside of the moving window average by ± 1.5
 26 standard deviations is removed. While the Bucsele et al. (2013) algorithm uses the same
 27 window size in a boxcar filtering step, it is performed later and only remove values above the
 28 mean (“hotspots”). Here, we perform this boxcar filter in both directions (above and below
 29 the mean) to remove anomalously low values that might result from a biased a-priori
 30 tropospheric estimate that was not accounted for in the masking step (avoiding negative

1 stratospheric NO₂ values being retained in subsequent steps), and to remove anomalously
2 high values that might result from transient pollution events that were likewise missed in the
3 masking step. We perform this boxcar filter twice to strictly remove outliers from regions
4 with noisy data.

5 Missing bins are then interpolated using a 30° longitude x 20° latitude moving window.
6 We tested smaller window sizes and found that they could introduce unphysical variability,
7 and/or leave missing data. Figure 1e shows how all the missing data over the TEMPO domain
8 are successfully filled using this window size. A few remaining “hot spots” are accounted for
9 in a third pass of the boxcar filter.

10 To obtain our final stratospheric NO₂ column estimate, we apply a final simple
11 smoothing step with a 5° x 3° window, as in Bucsela et al. (2013). The smaller box-car
12 window size in this step recognizes, and allows for, some regional scale variability in the
13 stratosphere. Figure 1f shows the final stratospheric NO₂ column estimate over the TEMPO
14 field of regard. Variation is primarily a function of latitude, from around 2×10^{15} molec cm⁻²
15 at the lowest latitudes in the field of regard (~20° latitude) to around 4×10^{15} molec cm⁻² at
16 the highest latitudes (~60° latitude). It is also apparent that this spatial filtering algorithm
17 allows for important regional scale variability to be retained in the stratospheric estimate.

18 In an effort to evaluate our new TEMPO algorithm with an independent estimate, we
19 compare our stratospheric vertical column with the stratospheric vertical column included in
20 the OMI SPv3 retrieval. Despite using different prior tropospheric estimates, incorporating
21 observations from GOME-2 outside the field of regard during interpolation, and employing
22 different box-car filtering steps, our algorithm is highly consistent with the results from the
23 global NASA standard OMI product over the TEMPO field of regard ($r = 0.972$, $m = 0.986$).
24 Overall, we calculate a mean bias in our new TEMPO algorithm compared to the NASA
25 standard product of only -0.05×10^{15} molecules cm⁻² (a normalized mean bias of -1.5 %).

26 Figure 2 shows the results of the same algorithm from an example day in January. The
27 shape of the expected TEMPO domain is impacted by large solar zenith angles at the highest
28 latitudes (we again use a solar zenith angle cut-off of 80°). Tropospheric enhancements
29 feature more prominently in the total slant column (Figure 2a) than in July since stratospheric
30 NO₂ columns are lower in the winter, and tropospheric NO₂ columns are higher. Figure 2b
31 shows the initial stratospheric estimate (V_{init}) from Equation 1, again using the monthly mean
32 GOME-2 tropospheric NO₂ column as an a priori estimate (Equation 2). Figure 2c shows the

1 result of applying the masking threshold (Equation 3). We find this threshold removes 51% of
2 the available data on average for this month (~21% of the available data are removed in the
3 global algorithm in January). Over the TEMPO domain we find that a slightly smaller fraction
4 pixels are removed in January compared to July because, despite having generally higher NO₂
5 tropospheric column densities, tropospheric air mass factors across the northeast are
6 extremely low at this time of year (discussed below). The low values are primarily due to
7 increased wintertime cloudiness. In this case, the masking threshold did not remove a strong
8 enhancement over the center of the continent. This highlights some criticism by Beirle et al.
9 (2016) of spatial filtering algorithms that rely strongly on a-priori climatologies wherein
10 transient tropospheric events could be misinterpreted as stratospheric. We find that varying
11 the magnitude of the threshold (Equation 3) does not successfully correct for this, since our
12 masking approach is based on a monthly mean and does not identify transient events, but this
13 feature is diminished in subsequent steps. Figure 2d shows the estimated stratospheric NO₂
14 outside of the TEMPO field of regard from the independent GOME-2 observations. Again,
15 these LEO observations provide powerful context despite being from a different time of day.
16 Figure 2e shows the result of the first two passes of the boxcar filter, and interpolating
17 unfilled bins using the 30° longitude x 20° latitude moving window.

18 Figure 2f shows the final stratospheric NO₂ estimate after the final pass of the statistical
19 test and 5° x 3° smoothing. The large enhancement of NO₂ over the continent has been
20 substantially dampened by our statistical filtering. The variability in the stratospheric NO₂
21 column is again generally latitudinal as expected, with values above 2×10^{15} molec cm⁻² at
22 the low latitudes, and below 1×10^{15} molec cm⁻² at the high latitudes.

23 The full TEMPO domain will have simultaneous sunlit coverage from about 1400 UTC
24 to 2300 UTC in July, and for only a few hours in January, based on a solar zenith angle
25 threshold of ~80°. Of concern is the lack of coverage over the west coast in the morning, and
26 over the east coast in the evening, where sunlit observations will not be available. Under these
27 circumstances, the stratospheric separation algorithm is challenged by even narrower spatial
28 domains. We evaluate these cases by repeating the calculations at specific times of day.

29 Figure 3 shows how the TEMPO algorithm would operate for 1130 Coordinated
30 Universal Time (UTC), 6:30 a.m. Eastern Standard Time (EST), on the example day in July.
31 Daylight observations over eastern North America are available by this time, without
32 coverage over the rest of the continent. All the algorithm steps are identical to those in Figure

1 1 and Figure 2 other than treatment of this partial coverage (additional near-real-time
2 considerations are discussed in Section 5). Figure 3a shows the OMI total slant columns. By
3 6:30 a.m. EST TEMPO observes only eastern North America. The availability of observations
4 increases in width northward because of the TEMPO viewing geometry. Figures 3b and 3c
5 show the initial stratospheric estimate (according to Equation 1) and the masked stratospheric
6 estimate (according to Equation 3) respectively. Figure 3d shows the independent LEO
7 observations from GOME-2 outside of the TEMPO field of regard. The observations are
8 binned, pass the statistical filtering steps, and interpolated in Figure 3e. The final stratospheric
9 estimate is shown in Figure 3f. Comparing this final stratospheric NO₂ estimate with the
10 estimate in Figure 1f (where coverage over the whole continent is assumed to be available),
11 we see the reduced coverage has negligible impact the final stratospheric estimate, and
12 identical spatial features are preserved ($R^2 = 0.995$).

13 Likewise, Figure 4 shows how the algorithm would operate on the example day in
14 January at 2330 UTC, or 3:30 pm Pacific Standard Time (PST). In addition to the loss of
15 observations in the east due to the time of day, larger solar zenith angles in the north at this
16 time of year further diminish coverage. Again, the subsequent steps are otherwise identical to
17 those in Figures 1 through 3. Figure 4a shows the OMI total slant columns. Observations are
18 available over parts of the Pacific Northwest, with coverage widening southward so that
19 observations are available from California to the western edge of Texas, and over western
20 parts of Mexico. Figure 4b and 4c show the initial stratospheric estimate (according to
21 Equation 1) and the masked stratospheric estimate (according to Equation 3) respectively.
22 Figure 4d shows how the independent LEO observations from again GOME-2 provide
23 coverage outside of the TEMPO field of regard. After binning and interpolation (Figure 4e)
24 followed by hot spot removal and smoothing, the final TEMPO stratospheric estimate is
25 shown in Figure 4f. Comparing this stratospheric NO₂ estimate with Figure 2f (where
26 coverage over the whole continent is assumed to be available) demonstrates again how the
27 reduced coverage has negligible impact the final stratospheric estimate, and identical spatial
28 features are preserved ($R^2 = 0.997$).

29 Next, we examine in detail the potential information penalty associated with the limited
30 TEMPO field of regard compared to a global implementation of our algorithm, and
31 demonstrate quantitatively that our approach can produce a tropospheric NO₂ estimate that is
32 consistent with a global algorithm, regardless of the time of day.

1

2 **4 Stratosphere-Troposphere Separation over the TEMPO Field of Regard**

3 The final step in the algorithm is the subtraction of the stratospheric NO₂ estimate from
4 the total slant column to obtain the tropospheric NO₂ column by:

$$5 \quad V_{trop} = \frac{(S - V_{strat} \cdot A_{strat})}{A_{trop}} \quad \text{Equation 5}$$

6 For this calculation we use the stratospheric and tropospheric air mass factors provided with
7 the OMI data product (the operational TEMPO algorithm would use TEMPO air mass
8 factors).

9 The difference between two tropospheric NO₂ column retrievals ($V_{trop,2}$ and $V_{trop,1}$) that
10 result from two different stratospheric NO₂ estimates ($V_{strat,2}$ and $V_{strat,1}$), but identical slant
11 columns and air mass factors, is directly proportional to the ratio of the tropospheric to
12 stratospheric air mass factors:

$$13 \quad V_{trop,2} - V_{trop,1} = \frac{A_{strat}}{A_{trop}} (V_{strat,2} - V_{strat,1}) \quad \text{Equation 6}$$

14 This means that differences (or errors) in stratospheric NO₂ estimates are magnified in the
15 tropospheric NO₂ column depending on the local air mass factors. This issue is particularly
16 important over the eastern US in the winter, where tropospheric air mass factors can be very
17 low (<0.1), and stratospheric air mass factors can be high (~5) depending on viewing
18 geometry. Figure 5 shows the stratospheric and tropospheric air mass factors for January 15,
19 2007. Over areas of the eastern US, where clouds prevail, the tropospheric air mass factors are
20 exceedingly small (~0.01), which gives rise to extremely large A_{strat}/A_{trop} ratios (>200). In
21 other words, residuals between two stratospheric NO₂ algorithms can become magnified by
22 more than two orders of magnitude in the troposphere.

23 The impact of errors in the tropospheric column due to this issue can be minimized by
24 excluding observations with high stratospheric to tropospheric air mass factor ratios. This is
25 also based on the logic that such values indicate tropospheric NO₂ is making a small
26 contribution to the measured signal (and as a result, the tropospheric NO₂ retrieval should
27 have high uncertainty). For this reason, we restrict all tropospheric NO₂ estimates to where
28 the local stratospheric to tropospheric air mass factor ratios are less than 5.

1 Figure 6 shows the stratospheric and tropospheric NO₂ columns estimated for July 15,
2 2007. The top panels display the stratospheric and tropospheric NO₂ columns as derived from
3 our TEMPO algorithm that employs the OMI data as a surrogate for TEMPO observations,
4 with adjacent GOME-2 data provided context outside the field of regard. The middle panels
5 display the stratospheric and tropospheric columns derived from implementing our algorithm
6 globally with OMI data alone (the results are restricted to the TEMPO field of regard in the
7 figure to facilitate comparison). The bottom panel shows the differences between our TEMPO
8 algorithm and the global algorithm. We find excellent spatial agreement in the tropospheric
9 NO₂ estimate between the two algorithms ($R^2 = 0.997$, slope = 1.008). More than 95% of the
10 pixels have differences that are smaller than $\pm 0.1 \times 10^{15}$ molec cm⁻².

11 We further evaluate the performance of our algorithm by comparing the tropospheric
12 NO₂ column distribution along the western-most edge (1° deep) of the TEMPO field of regard
13 with the tropospheric NO₂ tropospheric column distribution included in the independent
14 NASA SPv3 retrieval. In this relatively remote region of the field of regard, we find a similar
15 mean and standard deviation in column density ($0.71 \times 10^{14} \pm 3.63 \times 10^{14}$ molec cm⁻² in our
16 TEMPO algorithm and $0.98 \times 10^{14} \pm 3.38 \times 10^{14}$ molec cm⁻² in the NASA SPv3). The fraction
17 of negative columns that are observed in our algorithm is consistent with the fraction of
18 negative columns that occurs at the same location from the standard product (~37%).

19 Figure 7 compares the stratospheric and tropospheric NO₂ column estimates from the
20 TEMPO and global algorithms for January 15, 2007. The loss of coverage in the troposphere
21 (mostly over the eastern US) is a result of the air mass factor issue discussed above, leading to
22 tropospheric NO₂ retrievals with low information content. The spatial agreement in the
23 tropospheric NO₂ estimates that remain is excellent across the domain ($R^2 = 0.996$ slope =
24 0.999). The magnitude of the differences in the stratospheric columns become larger in the
25 troposphere, exceeding 0.5×10^{15} molec cm⁻² near the edges. Nonetheless, ~95% of the pixels
26 are consistent with the global version of the algorithm to within 0.25×10^{15} molec cm⁻².

27 Figure 8 shows the monthly mean tropospheric NO₂ columns resulting from our TEMPO
28 stratosphere-troposphere separation algorithm for both July and January, and the difference
29 versus results from the global algorithm. We find that our TEMPO algorithm produces
30 monthly mean results with negligible difference compared to the global algorithm, even at the
31 field of regard edges. The correlation between the two algorithms is excellent ($R^2 = 0.999$ and
32 slope = 1.009 for July, $R^2 = 0.998$ and slope = 0.999 for January). For July, more than 99% of

1 the pixels have differences that are smaller than $\pm 0.05 \times 10^{15}$ molec cm^{-2} . For January, more
2 than 90% of the pixels have differences that are smaller than $\pm 0.05 \times 10^{15}$ molec cm^{-2} , and
3 more than 99% of the pixels have differences that are smaller than $\pm 0.10 \times 10^{15}$ molec cm^{-2} . In
4 other words, our TEMPO-specific algorithm performs almost identically to the LEO
5 algorithm that uses all available global data. There are some random errors near the field of
6 regard edges on individual days (Figures 6 and 7), but these nearly disappear in the monthly
7 average (Figure 8)

8 Figure 9 shows the July monthly mean tropospheric NO_2 columns resulting from
9 retrievals at 1130 UTC (east coast summer morning) and at 0200 UTC (west coast summer
10 evening). The east coast morning retrieval example exhibits small positive biases over some
11 the Great Lakes region compared to the global algorithm, but overall the spatial agreement
12 remains excellent ($R^2 = 0.996$ and slope = 1.015). More than 90% of the pixels have
13 differences that are smaller than $\pm 0.05 \times 10^{15}$ molec cm^{-2} , and more than 98% of the pixels
14 have differences that are smaller than $\pm 0.10 \times 10^{15}$ molec cm^{-2} . The west coast summer
15 evening example also exhibits excellent performance overall ($R^2 = 0.998$ and slope = 0.994).
16 In this case, more than 98% of the pixels have differences that are smaller than $\pm 0.05 \times 10^{15}$
17 molec cm^{-2} .

18 Figure 10 shows the January monthly mean tropospheric NO_2 columns resulting from
19 retrievals at 1400 UTC (east coast winter morning) and 2330 UTC (west coast winter
20 evening). The bottom panels in Figure 10 show the difference between the results from our
21 TEMPO algorithm and the results from the global algorithm. In the east coast winter case,
22 spatial agreement is still very good in general ($R^2 = 0.995$), but we find noticeable
23 degradation in the absolute performance over the continent compared to the global algorithm
24 resulting from this partial field of view (slope = 1.038). The west coast winter evening
25 retrieval performs better overall ($R^2 = 0.999$, slope = 1.007). Although the algorithm performs
26 poorest in the east coast winter morning case, $\sim 90\%$ of the tropospheric pixels still have
27 differences that are less than 0.2×10^{15} molec cm^{-2} , a commonly accepted estimate of the
28 stratospheric uncertainty resulting from stratosphere-troposphere separation in NO_2 retrieval
29 algorithms (Boersma et al., 2004). Moreover, two hours later at 1600 UTC when the field of
30 regard has expanded across the Great Lakes region, into the middle of North America, and
31 covers most of Mexico, this issue disappears ($R^2 = 0.999$, slope = 0.998). In other words, as
32 spatial coverage expands, the absolute constraint on stratospheric NO_2 becomes more robust.

1 This highlights the challenge of accurate wintertime tropospheric NO₂ retrievals
2 (especially over eastern North America) when pollution is primarily in a shallow boundary
3 layer close to the surface where satellite remote sensing sensitivity is lowest. The partial
4 TEMPO field of regard in this case exacerbates the problem, but the challenge is not unique
5 to TEMPO retrievals.

6 Finally, we further test the performance of this algorithm at other times of day by
7 repeating the same steps as above, but using GOME-2 observations as a surrogate for
8 TEMPO. For this, we swap all instances of the OMI observations (overpass time ~ 13:30)
9 with GOME-2 observations (overpass time ~09:30), and vice versa. In other words, the
10 GOME-2 observations are restricted to the anticipated field of regard, and we use a monthly
11 from OMI as our a priori tropospheric column and the daily observations from OMI as
12 supporting global observations outside the TEMPO field of regard. We find the performance
13 at this morning overpass time is as good as the mid-afternoon overpass time ($R^2 = 0.999$,
14 slope = 1.005 for July; and $R^2 = 0.999$, slope = 1.005 for January), providing more evidence
15 that our approach works equally well at different times of day.

16

17 **5 Near-Real-Time Considerations**

18 For retrievals in near-real time (i.e. within an hour of the observation), independent
19 global observations in LEO may not be available (e.g. unexpected issues with LEO
20 observation processing). Here we test the performance of the TEMPO algorithm without the
21 supporting global observations by carrying out the identical steps outlined in Sections 3 and 4
22 except without incorporating the GOME-2 observations outside the TEMPO field of regard.
23 Comparing these results with the global algorithm isolates the penalty due to the limited
24 TEMPO spatial domain alone, since the steps are otherwise computationally identical.

25 Figure 11 shows the mean July and January tropospheric columns resulting from this
26 near-real time test. The spatial correlation with the global algorithm is still strong overall (R^2
27 = 0.924 and slope = 0.973 for July and $R^2 = 0.996$ and slope = 1.008 for January), and
28 between 90-95% of pixels in both July and January differ from the global algorithm by less
29 than 0.2×10^{15} molec cm⁻². We find that, compared to a global algorithm, this stratosphere-
30 troposphere separation approach gives rise to noticeable systematic biases near the field of
31 regard edges (including Mexico, the Caribbean, and northern Canada). The differences are
32 due to the lack of supporting data outside of the TEMPO field of regard.

1 This is most evidently a problem near the northern/southern borders of the field of regard,
2 given the strong gradient in stratospheric NO₂ as a function of latitude. At low latitudes, when
3 the averaging windows intersect with the field of regard, the global algorithm would have
4 lower mean values by including observations to the south. This causes the stratospheric
5 column from the TEMPO algorithm to be systematically biased high compared to the global
6 algorithm, translating into an underestimate in the tropospheric column (by more than $-0.5 \times$
7 10^{15} molec cm⁻² in some locations). By the same logic, there is a high bias (also more than
8 $+0.5 \times 10^{15}$ molec cm⁻² on average) along the northern edge of the field of regard in July.
9 There are also small low biases in the tropospheric column throughout the eastern side of the
10 TEMPO field of regard over the Atlantic Ocean. By excluding more pristine ocean conditions
11 further to the east, the stratospheric column derived by the TEMPO algorithm is biased high
12 compared to the global algorithm, which again translates into an underestimate in the
13 tropospheric column.

14 In the absence of daily ancillary satellite data for estimating stratospheric NO₂ outside the
15 field of regard, a climatology built from satellite observations or model data could mitigate
16 these edge effects for near real time retrievals since the average latitudinal and seasonal
17 dependence of stratospheric NO₂ are generally well known. For example, tests conducted
18 using a monthly mean global stratospheric NO₂ estimate as the supporting data outside the
19 TEMPO field of regard improves the correlations in both cases ($R^2 = 0.999$ and slope = 1.010
20 for July and $R^2 = 0.999$ and slope = 1.002 for January), now with >99% of the monthly mean
21 pixels differing from the global algorithm results by less than 0.05×10^{15} molec cm⁻².

22 Similarly, we find weaker overall performance in the cases of partial fields of regard
23 without context from surrounding LEO observations. Figure 12 shows the July mean
24 tropospheric column retrievals calculated for 1130 UTC (east coast summer morning) and the
25 July mean tropospheric column retrievals for 0200 UTC (west coast summer evening).
26 Though this version of the algorithm performs less well compared to the results from
27 incorporating independent LEO observations, the spatial correlation is still good ($R^2 = 0.944$,
28 slope = 0.943 for 1130 UTC July; $R^2 = 0.964$, slope = 0.986 for 0200 UTC). The differences
29 over most of the available domain remain small, with 90-95% of the pixels having differences
30 in the mean tropospheric column of less than $\pm 0.2 \times 10^{15}$ molec cm⁻² compared to the global
31 algorithm. Figure 13 shows the January mean tropospheric column retrievals calculated for
32 1400 UTC (east coast winter morning) and the January mean tropospheric column retrievals

1 for 2300 UTC (west coast winter evening). The spatial correlation in both cases remains
2 strong, again with some systematic biases observed ($R^2 = 0.996$, slope = 1.001 at 1400 UTC
3 and $R^2 = 0.987$, slope = 1.019 at 2330 UTC). The biases remain modest, with ~90% of the
4 pixels being consistent to within $0.2 \times 10^{15} \text{ cm}^{-2}$ of the global implementation of the
5 algorithm. Again, using a monthly climatology mitigates the biases in all cases, with the
6 smallest improvement for the retrieval in January at 1400 UTC (going from 90% to 94% of
7 the pixels being consistent to within $0.2 \times 10^{15} \text{ cm}^{-2}$ of the global implementation of the
8 algorithm).

9 Given these results, our recommendation for TEMPO is to use a climatological estimate
10 (e.g. a 30-day mean) of stratospheric NO_2 for context outside of the TEMPO field of regard
11 during near-real-time retrieval if LEO observations are unavailable. This climatological
12 estimate can be constructed based on satellite-derived observations in LEO from the
13 preceding year and corrected for the time of day based on model results or other independent
14 observations. We would then propose a subsequent re-processing of the data that incorporates
15 the daily LEO observations when available from the correct observation day.

16

17 **6 Conclusions**

18 The TEMPO geostationary satellite instrument is expected to provide hourly observations
19 of NO_2 columns (among a variety of other measurements) over North America. Here, we have
20 developed and tested the first stratosphere-troposphere separation algorithm for TEMPO
21 geostationary satellite observations of atmospheric NO_2 column density. We use independent
22 measurements from a low Earth observing satellite instrument to identify likely locations of
23 tropospheric enhancements, and to provide context outside of the available TEMPO
24 measurements. We consider partial fields of regard as a function of time of day, and
25 implement a new filter based on stratospheric to tropospheric air mass factor ratios. We
26 investigate in particular the information penalty associated with the limited TEMPO fields of
27 regard as a function of season and time of day.

28 We find that our algorithm performs as well as a global LEO algorithm for most
29 scenarios. When the whole continent is observed, monthly mean agreement with tropospheric
30 NO_2 retrieved from the global algorithm is excellent ($R^2 = 0.999$, slope = 1.009 for July and
31 $R^2 = 0.998$, slope = 0.999 January). During most instances with a partial field of regard (e.g.
32 east coast morning or west coast evening) the algorithm still performs robustly. We

1 demonstrate that small biases near the southern and northern edges of the field of regard are
2 avoided by incorporating independent LEO observations that have been corrected for the time
3 of day. When the whole continent is observed, the vast majority of pixels ($> 95\%$) agree with
4 results from a global implementation of the same algorithm to within $\pm 0.05 \times 10^{15}$ molec cm^{-2} .
5 We find that the TEMPO algorithm is challenged most by winter east coast morning
6 retrievals, but nonetheless the difference between the TEMPO algorithm and the global
7 implementation of the same algorithm produces differences that are less than 0.2×10^{15} molec
8 cm^{-2} for more than 90% of the pixels. Even when supporting observations from LEO may not
9 be available (as in near-real-time), a large majority of pixels ($\sim 90\%$ or greater) agree with the
10 global algorithm to within $\pm 0.2 \times 10^{15}$ molecules cm^{-2} on a monthly mean basis, which is
11 generally accepted as typical estimates of stratospheric error due to stratosphere-troposphere
12 separation algorithms. The differences can be reduced further in near-real-time retrievals by
13 the use of a climatology outside the TEMPO field of regard. The value of independent LEO
14 observations for TEMPO tropospheric retrievals implies benefit to TEMPO data from
15 ongoing development of LEO observations.

16 We have demonstrated a feasible and robust stratosphere-troposphere separation
17 algorithm for the retrieval of geostationary satellite-based NO_2 tropospheric column densities
18 by the TEMPO instrument notwithstanding the limited field of regard or changing time of
19 day. Our TEMPO algorithm also demonstrates good performance when evaluated against the
20 stratospheric NO_2 columns provided with the NASA SPv3 standard product, but further
21 independent evaluation using ground-based spectrometer network observations will be
22 beneficial. This approach may be applicable to other planned geostationary satellite
23 instruments including Sentinel-4 over Europe and GEMS over Asia. This spatial filtering and
24 interpolation method may also have applications in offset removal during retrievals of HCHO
25 and SO_2 tropospheric columns.

26

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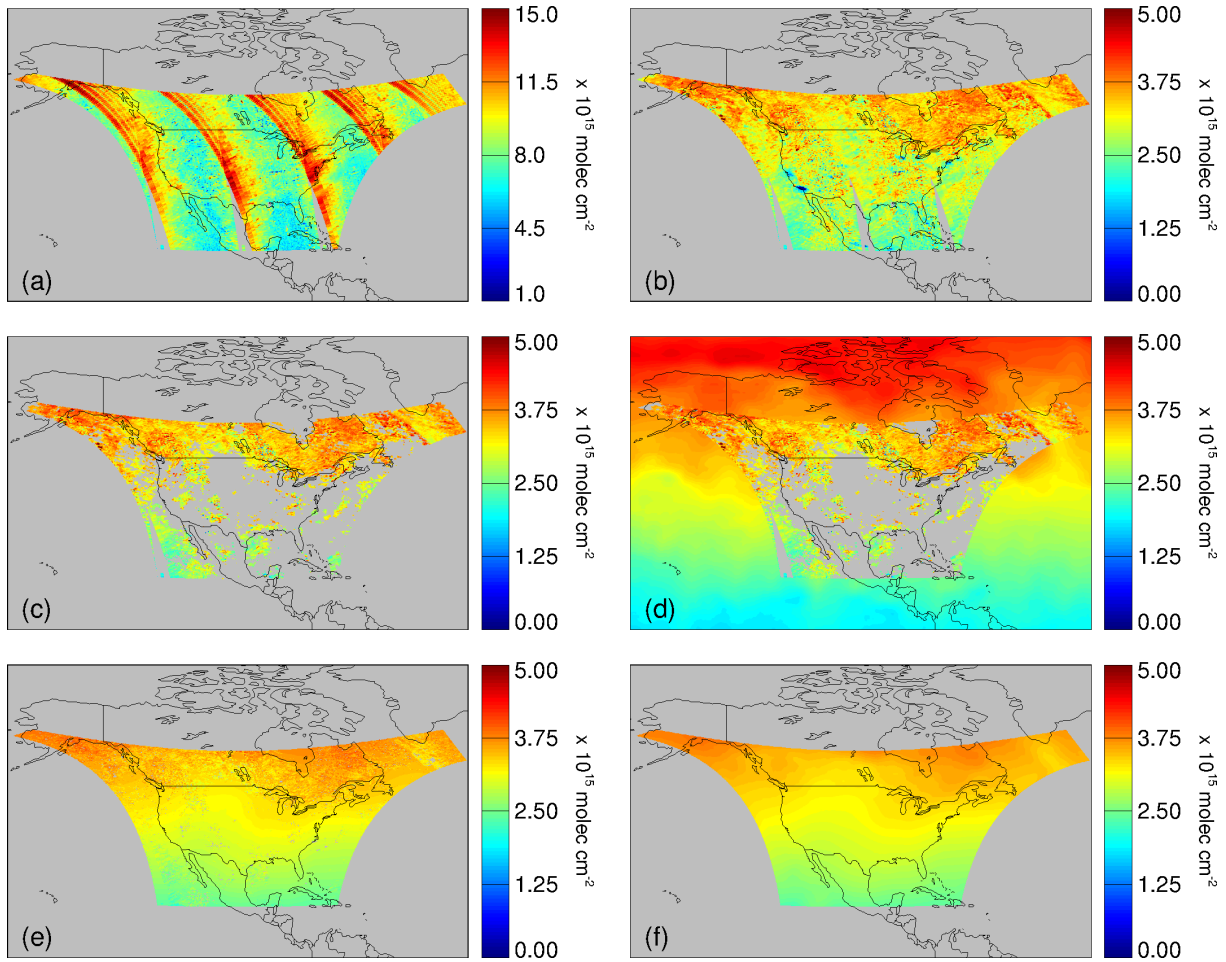
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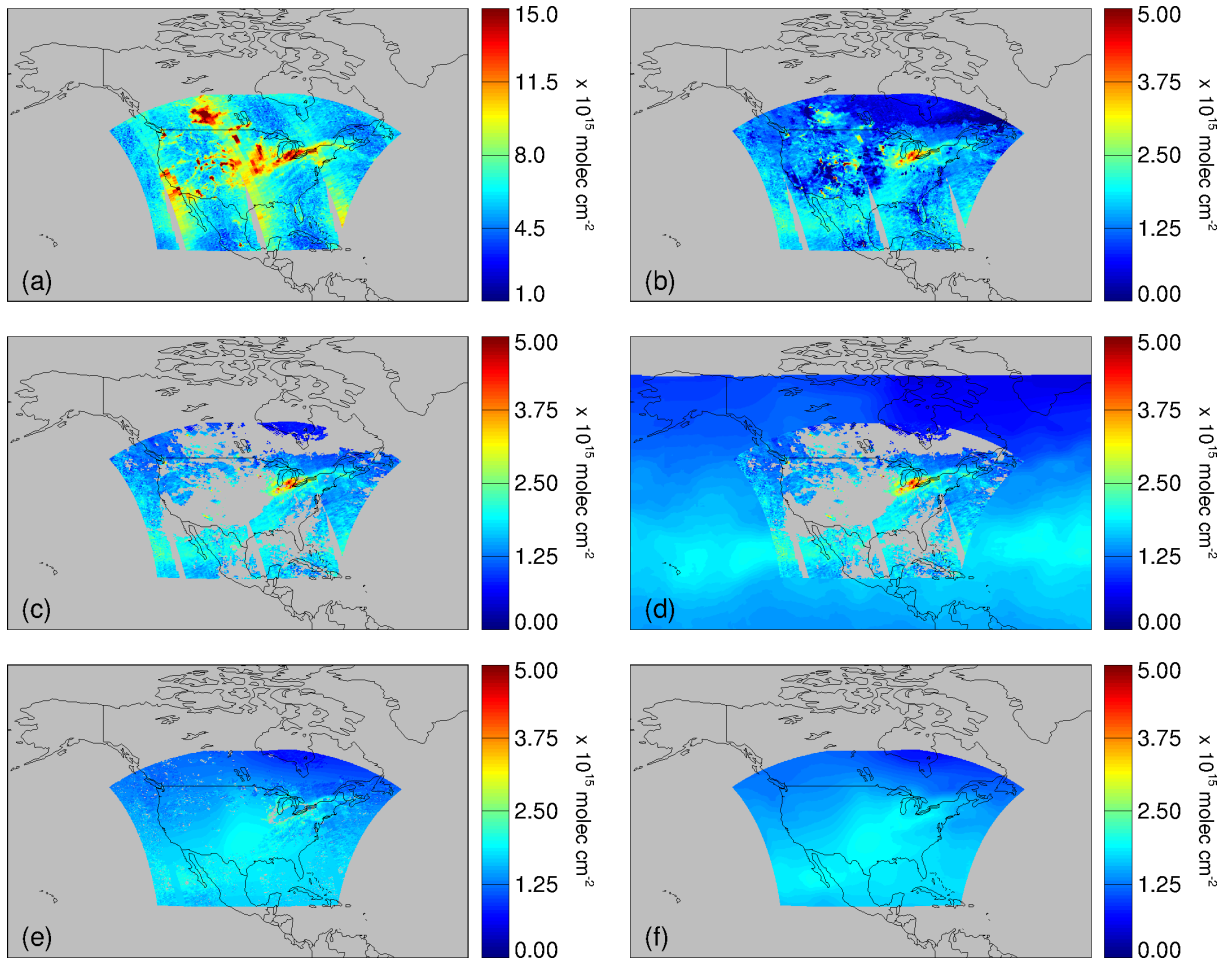
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2 **Figure 1:** Calculation of the stratospheric NO₂ estimate on July 15, 2007 using OMI
3 observations from within the anticipated TEMPO field of regard: (a) Slant columns on a 0.1°
4 x 0.1° grid. (b) Initial stratospheric estimate (V_{init}) resulting from Equation 1 and 2. (c)
5 Masked V_{init} using a threshold of $S_{trop}/A_{strat} < 0.3 \times 10^{15} \text{ molec cm}^{-2}$ to remove large
6 tropospheric influence. (d) Adding context outside of the TEMPO field of regard by using
7 independent low-earth orbit observations from GOME-2 that have been corrected for time of
8 day. (e) Stratospheric NO₂ estimate with masked areas interpolated. (f) Stratospheric NO₂
9 estimate after final hot spot removal and smoothing.

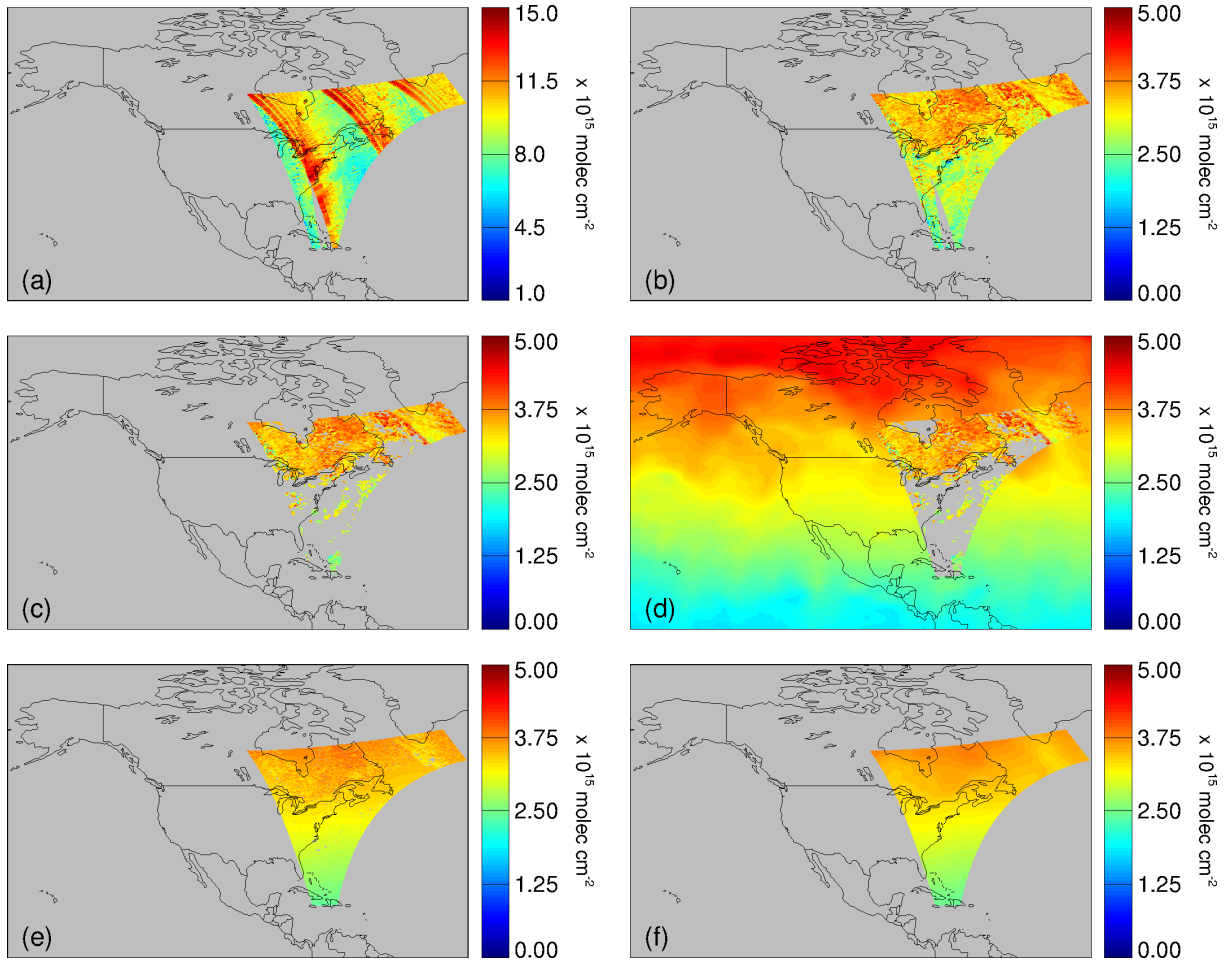
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2 **Figure 2:** Calculation of the stratospheric NO₂ estimate on January 15, 2007 using OMI
3 observations from within the anticipated TEMPO field of regard: (a) Slant columns on a 0.1°
4 x 0.1° grid. (b) Initial stratospheric estimate (V_{init}) resulting from Equation 1 and 2. (c)
5 Masked V_{init} using a threshold of $S_{trop}/A_{strat} < 0.3 \times 10^{15} \text{ molec cm}^{-2}$ to remove large
6 tropospheric influence. (d) Adding context outside of the TEMPO field of regard by using
7 independent low-earth orbit observations from GOME-2 that have been corrected for time of
8 day. (e) Stratospheric NO₂ estimate with masked areas interpolated. (f) Stratospheric NO₂
9 estimate after final hot spot removal and smoothing.

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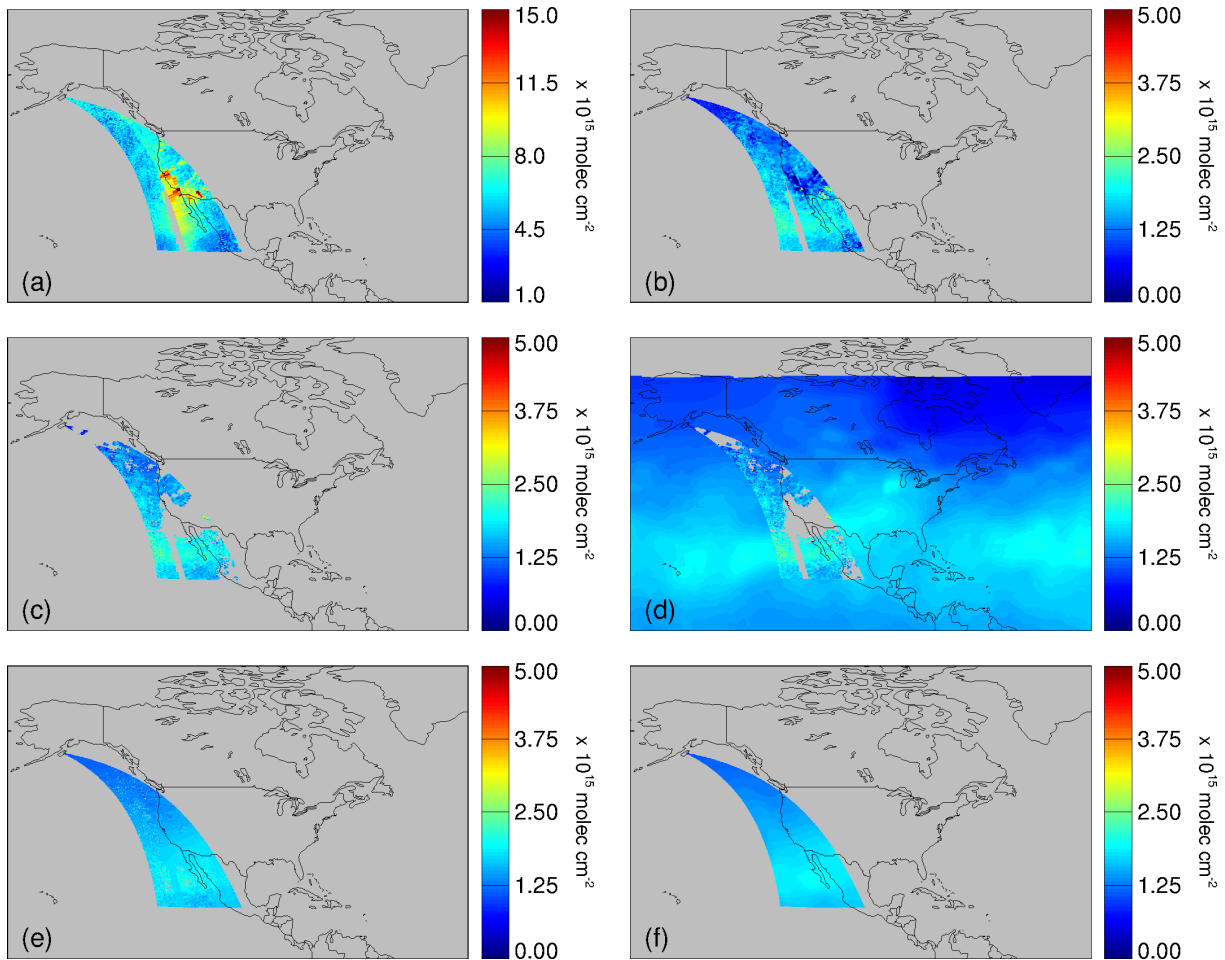


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3 **Figure 3:** Calculation of the stratospheric NO₂ estimate on July 15, 2007 using OMI
 4 observations from within the anticipated TEMPO field of regard at 1130 UTC (6:30 am
 5 Eastern Standard Time): (a) Slant columns on a 0.1° x 0.1° grid. (b) Initial stratospheric
 6 estimate (V_{init}) resulting from Equation 1 and 2. (c) Masked V_{init} using a threshold of S_{trop}/A_{strat}
 7 $< 0.3 \times 10^{15} \text{ molec cm}^{-2}$ to remove large tropospheric influence. (d) Adding context outside of
 8 the TEMPO field of regard by using independent low-earth orbit observations from GOME-2
 9 that have been corrected for time of day. (e) Stratospheric NO₂ estimate with masked areas
 10 interpolated and smoothed. (f) Stratospheric NO₂ estimate after final hot spot removal
 11 smoothing.

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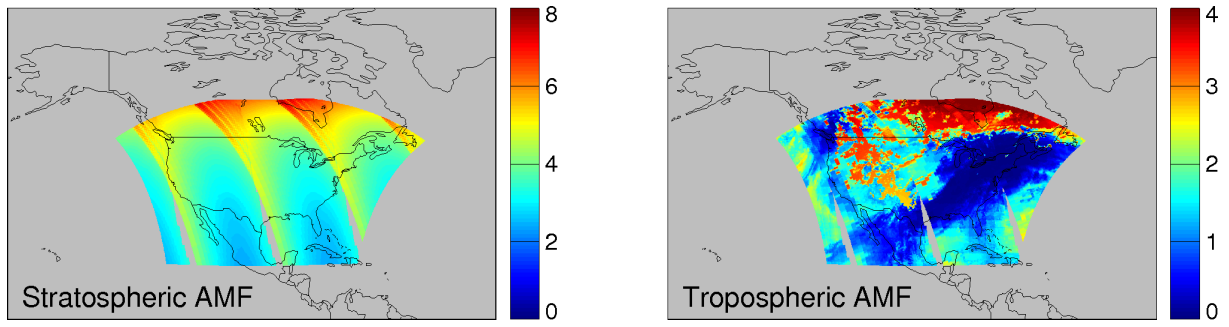
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3 **Figure 4:** Calculation of the stratospheric NO₂ estimate on January 15, 2007 using OMI
 4 observations from within the anticipated TEMPO field of regard at 2330 UTC (3:30 pm
 5 Pacific Standard Time): (a) Slant columns at 0.1° x 0.1° resolution. (b) Initial stratospheric
 6 estimate (V_{init}) resulting from Equation 2. (c) Masked V_{init} using a threshold of $S_{trop}/A_{strat} < 0.3$
 7 $\times 10^{15}$ molec cm⁻² to remove large tropospheric influence. (d) Adding context outside of the
 8 available TEMPO field of regard by using independent low-earth orbit observations from
 9 GOME-2 that have been corrected for time of day. (e) Stratospheric NO₂ estimate with
 10 masked areas interpolated and smoothed. (f) Final stratospheric NO₂ estimate after hot spot
 11 removal and smoothing.

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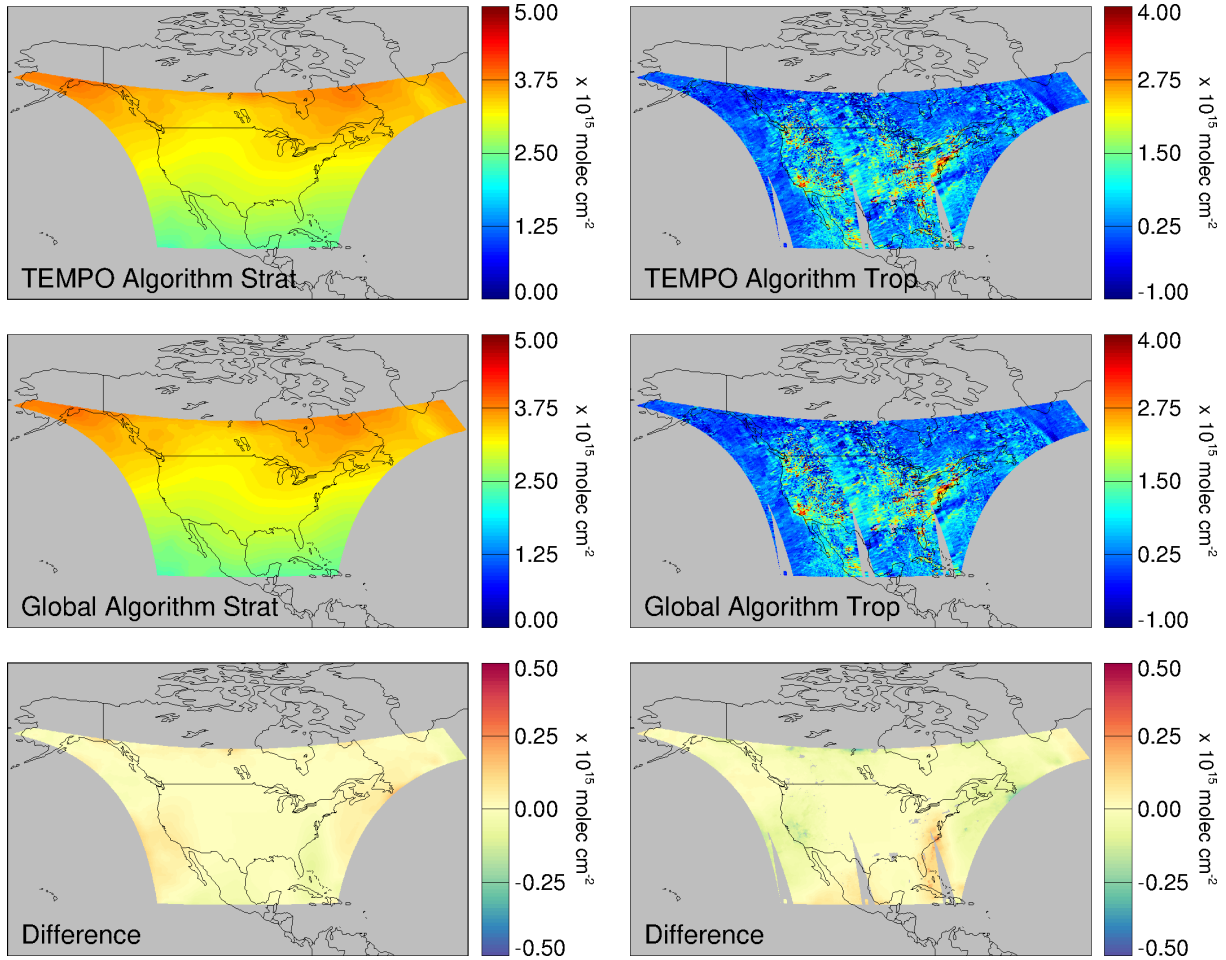


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Figure 5: Stratospheric (left) and tropospheric (right) air mass factors for January 15, 2007.

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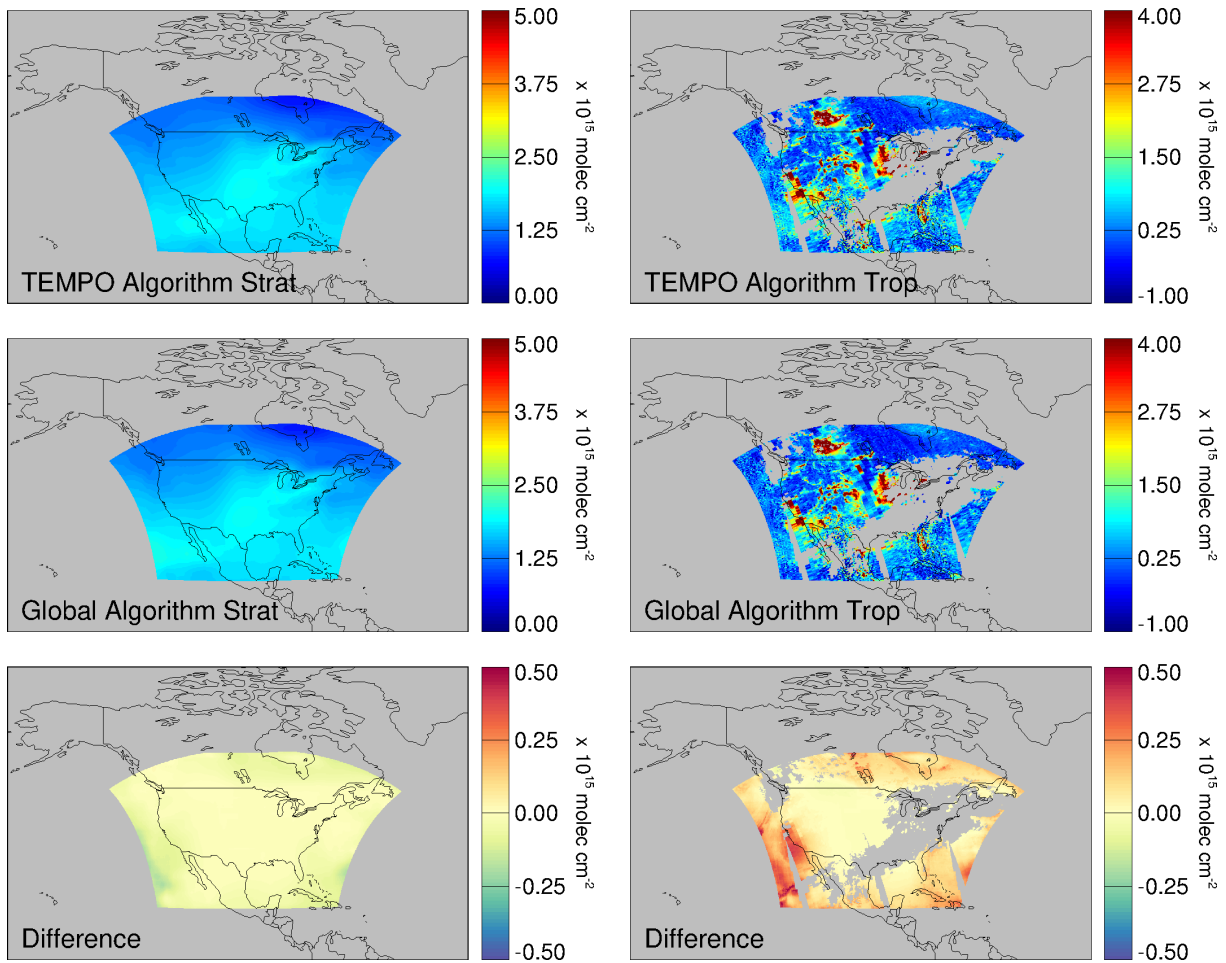


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3 **Figure 6:** Stratospheric NO₂ (left panels) and final tropospheric NO₂ retrievals (right panels)
4 resulting from our stratosphere-troposphere separation algorithms for July 15, 2007. Top
5 panels show the results using our proposed TEMPO algorithm. Middle panels show the
6 results using global observations (results have been clipped to the TEMPO field of regard for
7 comparison). Bottom panels show the absolute absolute differences between the TEMPO
8 and global algorithm results.

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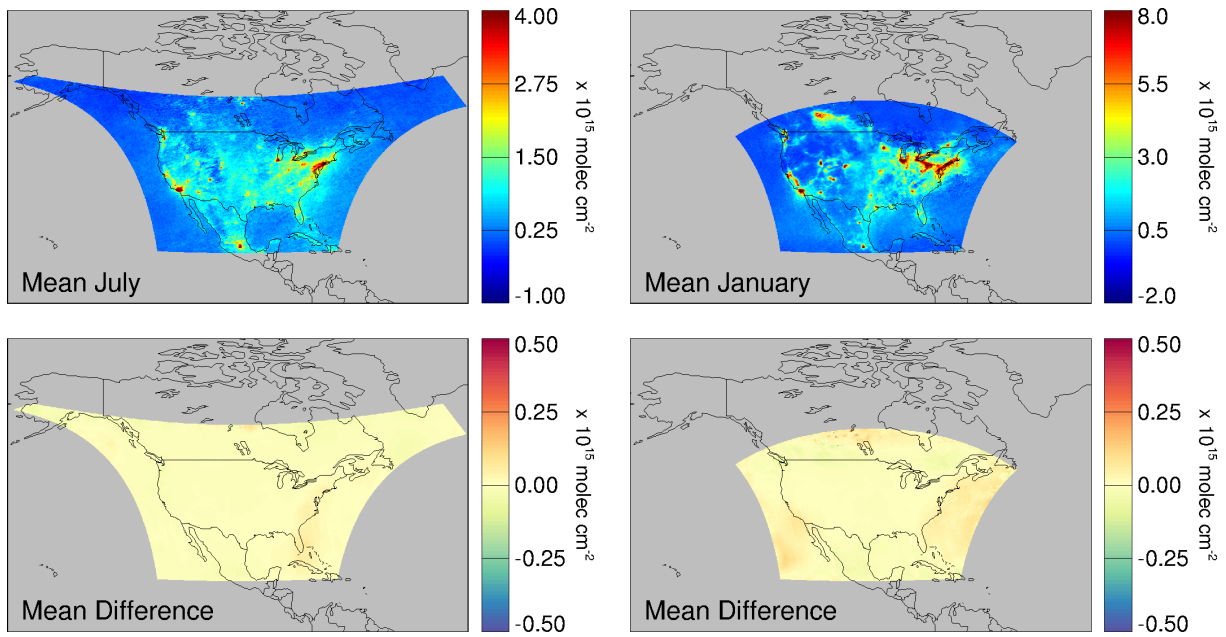


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3 **Figure 7:** Stratospheric NO₂ (left panels) and final tropospheric NO₂ retrievals (right panels)
4 resulting from our algorithm for January 15, 2007. Top panels show the results using our
5 proposed TEMPO algorithm. Middle panels show the results using global observations
6 (results have been clipped to the TEMPO field of regard for comparison). Bottom panels
7 show the absolute differences between the TEMPO and global algorithm results.

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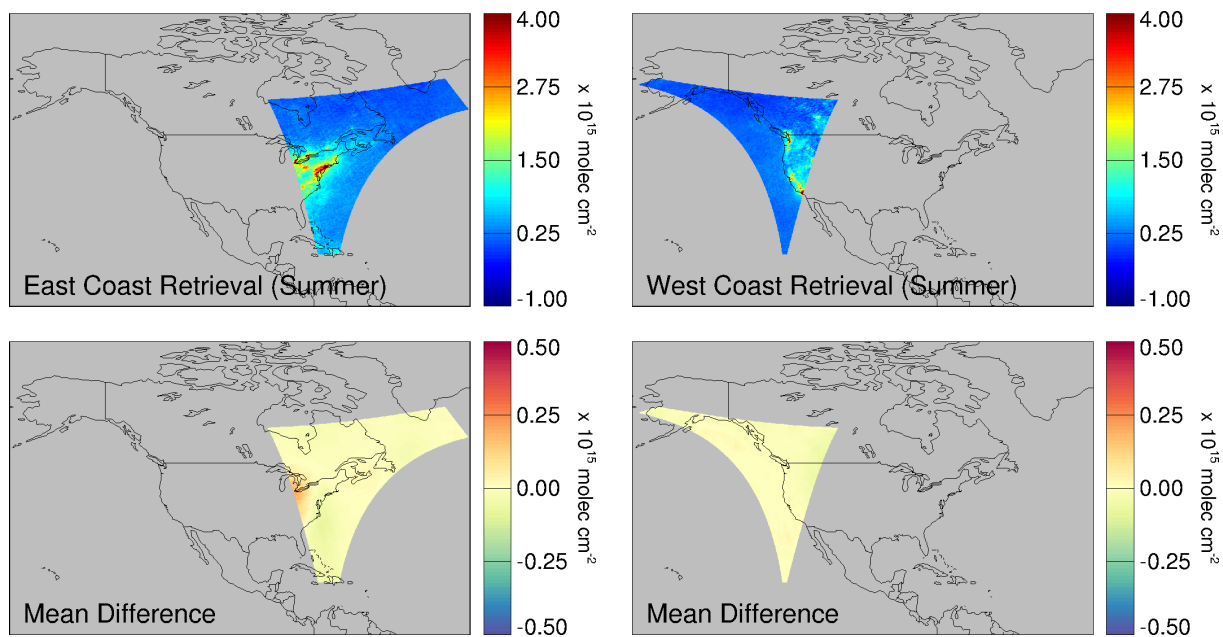


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3 **Figure 8:** Top panels show mean July and January tropospheric NO₂ column densities
4 resulting from our TEMPO algorithm. Bottom panels show absolute difference in mean July
5 and January tropospheric NO₂ between the TEMPO algorithm and the global algorithm.

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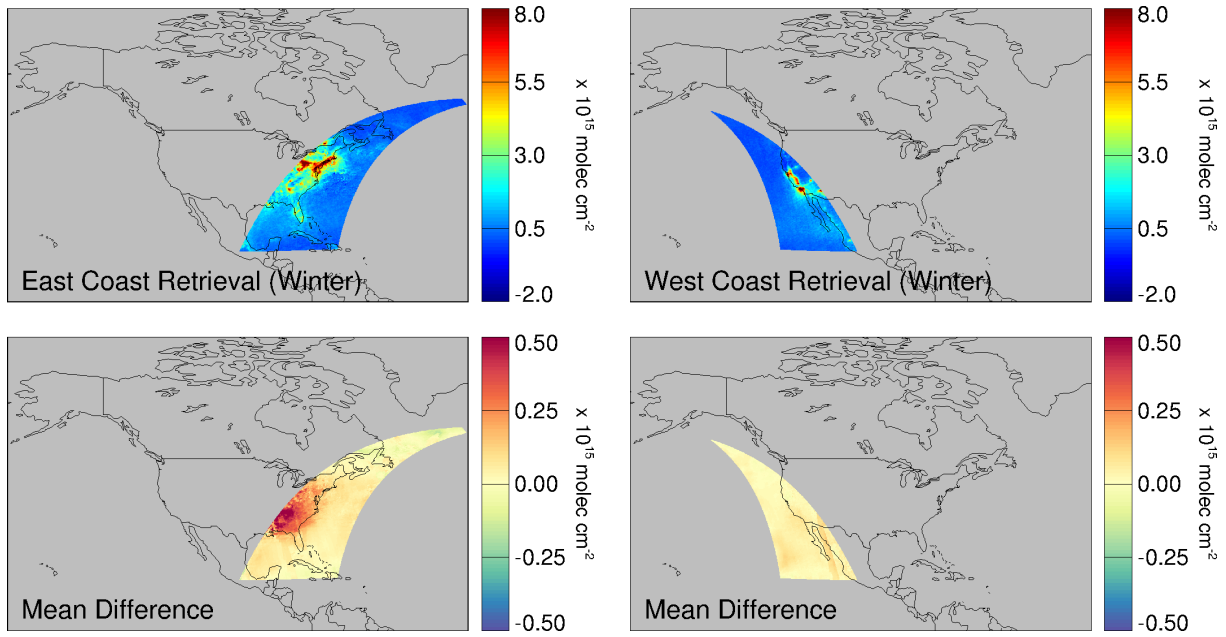


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3 **Figure 9:** Top panels show mean July tropospheric NO₂ column densities at 1130 UTC (left)
4 and 0200 UTC (right) resulting from our TEMPO STS algorithm. Bottom panels show
5 absolute difference in the tropospheric NO₂ column between the TEMPO algorithm and the
6 global STS algorithm.

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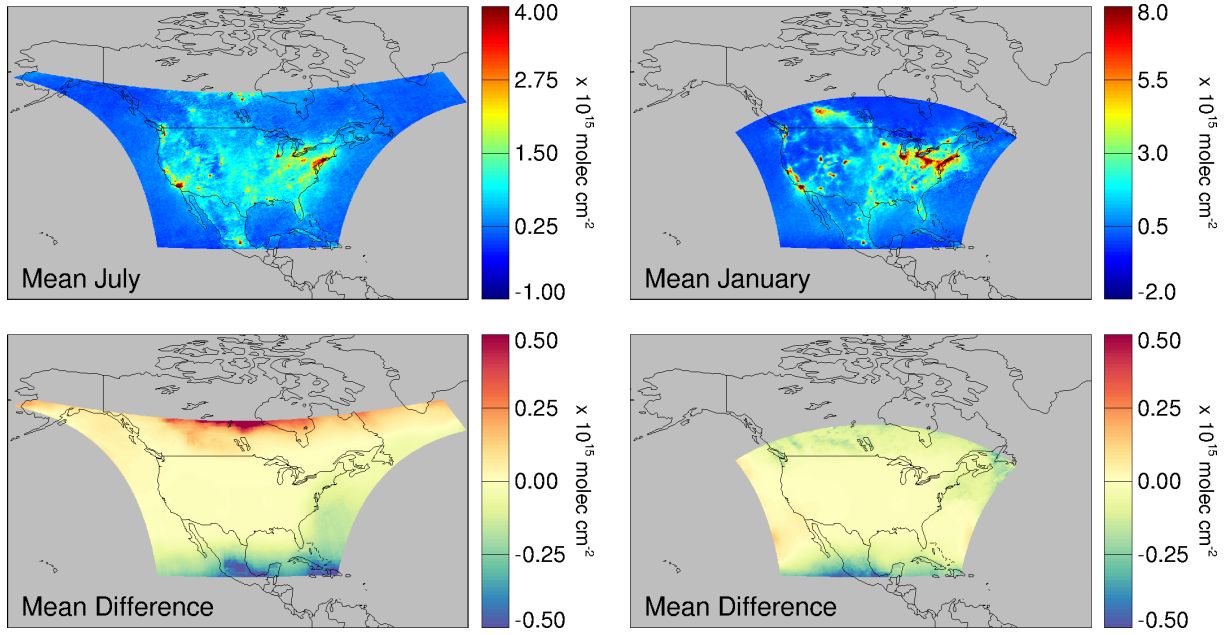


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3 **Figure 10:** Top panels show mean January tropospheric NO₂ column densities at 1400 UTC
4 (left) and 2330 UTC (right) resulting from our TEMPO STS algorithm. Middle panels show
5 absolute difference in the tropospheric NO₂ column between the TEMPO algorithm and the
6 global STS algorithm.

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3 **Figure 11:** Top panels show mean July and January tropospheric NO₂ column densities
4 resulting from our TEMPO STS algorithm without using independent low-earth orbit
5 observations for context outside the TEMPO field of regard (as might be occasionally
6 expected in near-real-time operations). Bottom panels show absolute difference in mean July
7 and January tropospheric NO₂ between the TEMPO algorithm and the global STS algorithm.

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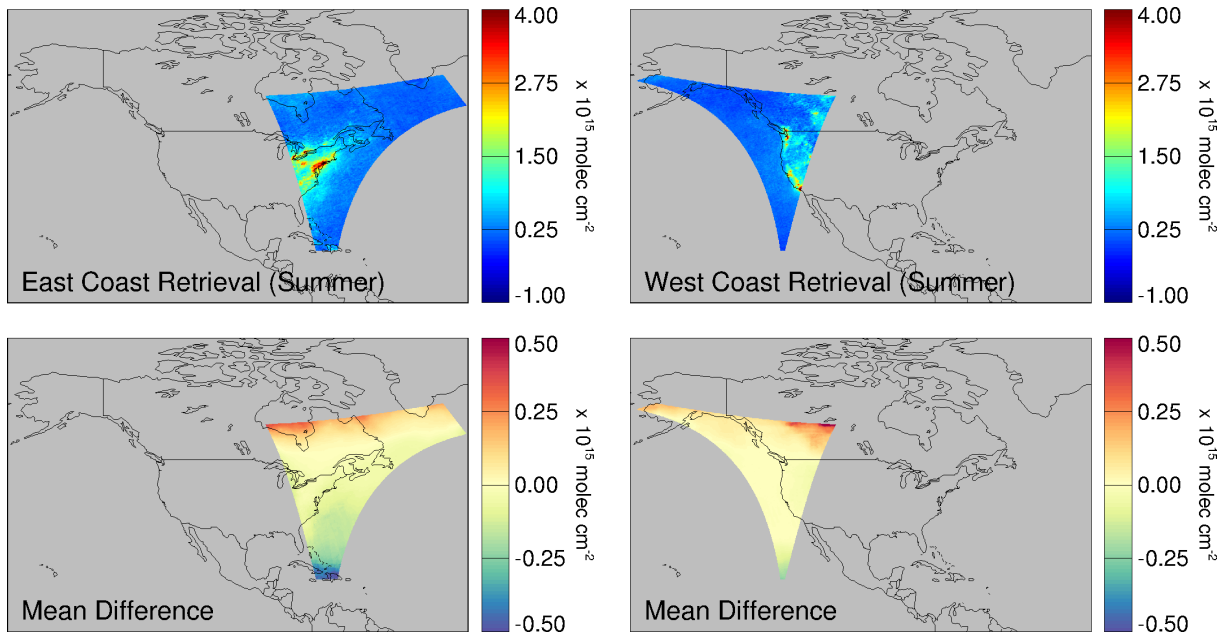
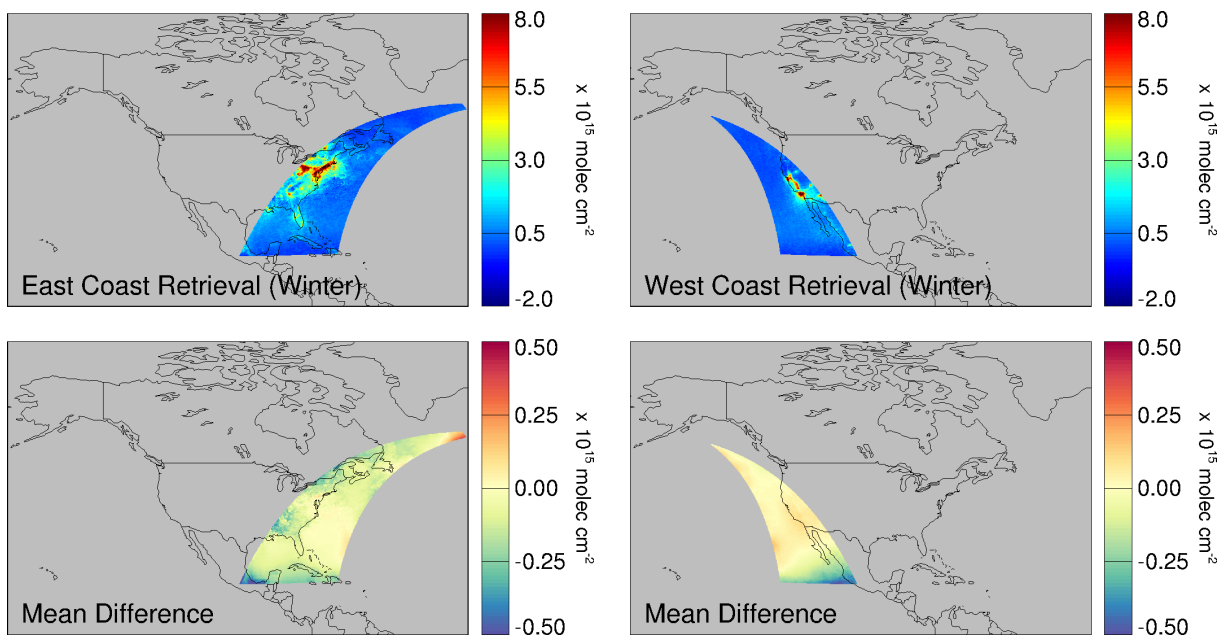


Figure 12: Top panels show mean July tropospheric NO₂ column densities at 1130 UTC (left) and 0200 UTC (right) resulting from our TEMPO STS algorithm without using independent low-earth orbit observations for context outside the TEMPO field of regard. Bottom panels show absolute difference in the tropospheric NO₂ column between the TEMPO algorithm and the global STS algorithm.

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3 **Figure 13:** Top panels show mean January tropospheric NO₂ column densities at 1400 UTC
4 (left) and 2330 UTC (right) resulting from our TEMPO algorithm without using independent
5 low-earth orbit observations for context outside the TEMPO field of regard. Middle panels
6 show absolute difference in the tropospheric NO₂ column between the TEMPO algorithm and
7 the global algorithm.

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