

A new TROPOMI product for tropospheric NO₂ columns over East Asia with explicit aerosol corrections

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Abstract. We present a new product with explicit aerosol corrections, POMINO-TROPOMI, for tropospheric nitrogen dioxide (NO₂) vertical column densities (VCDs) over East Asia, based on the newly launched TROPOspheric Monitoring Instrument with an unprecedented high horizontal resolution. Compared to the official TM5-MP-DOMINO (OFFLINE) product, POMINO-TROPOMI shows stronger concentration gradients near emission source locations and better agrees with MAX-DOAS measurements ($R^2 = 0.75$; NMB = 0.8 % versus $R^2 = 0.68$, NMB = -41.9 %). Sensitivity tests suggest that implicit aerosol corrections, as in TM5-MP-DOMINO, lead to underestimations of NO2 columns by about 25 % over the polluted northern East China region. Reducing the horizontal resolution of a priori NO₂ profiles would underestimate the retrieved NO₂ columns over isolated city clusters in western China by 35 % but with overestimates of more than 50 % over many offshore coastal areas. The effect of a priori NO₂ profiles is more important under calm conditions.

1 Introduction

Nitrogen oxides ($NO_x = NO + NO_2$) are crucial gaseous pollutants in the troposphere. NO_x leads to the production of particulate matter and ozone (O_3) and enhances levels of oxidants in the troposphere (Shindell et al., 2009), which affect air quality (Dentener et al., 2003) and human health (Hoek et al., 2013). Satellite remote sensing is widely used to monitor levels of nitrogen dioxide (NO_2) pollution worldwide (McLinden et al., 2014; Krotkov et al., 2016; Ott et al., 2010; Russell et al., 2011; Lin et al., 2010). The TRO- POspheric Monitoring Instrument (TROPOMI), which was jointly developed by the Netherlands and Europe Space Agency (ESA) (Veefkind et al., 2012) and was launched on 13 October 2017, is a UV–visible (Vis)–near-infrared– shortwave infrared backscattering sensor on board the sunsynchronous Sentinel-5 Precursor (S5P) satellite with an overpass time of 13:30 local solar time. With a wide swath of 2600 km and an unprecedentedly high horizontal resolution of $3.5 \text{ km} \times 7 \text{ km}$, ($3.5 \text{ km} \times 5.5 \text{ km}$ since 6 August 2019) TROPOMI achieves daily global coverage. This high horizontal resolution and good spatial coverage, combined with the very high signal-to-noise ratio, enables the instrument to resolve NO₂ pollution from point sources, medium-size urban centers, highways, or rivers, tasks that were difficult to achieve before.

Retrievals of tropospheric NO₂ vertical column densities (VCDs) in the UV-Vis spectral range from satellite instruments consist of three steps: (1) using differential optical absorption spectroscopy (DOAS) to fit the slant column density (SCD) of NO_2 along the light path, (2) subtracting the stratospheric contribution from the SCD to obtain the tropospheric SCD, and (3) converting the tropospheric SCD to the tropospheric VCD by using the calculated air mass factor (AMF). For TROPOMI, the random uncertainty in the total SCDs is $\sim 0.6 \times 10^{15}$ molec. cm⁻², considerably smaller than for the Ozone Monitoring Instrument (OMI; $\sim 0.8 \times 10^{15}$ molec. cm⁻²; Zara et al., 2018). The (total or stratospheric) bias is generally between 0% and -10% according to SAOZ observations (Eskes et al., 2019), which meets the error requirements as defined in the S5P Calibration and Validation Plan (Goryl et al., 2017). The calculation of the AMF introduces the dominant source of error in the retrieved tropospheric NO2 VCDs over polluted areas (Boersma et al., 2004, 2011, 2018; Lorente et al., 2017; Lin et al., 2014). The median negative biases of the daily comparisons between tropospheric VCDs from the Dutch official TM5-MP-DOMINO (OFFLINE) product and MAX-DOAS measurements are generally less than 50% (the allowable bias is 25 %-50 %; Goryl et al., 2017) but quite variable with the stations and NO₂ levels, especially at polluted locations (Eskes et al., 2018; van Geffen et al., 2019).

TM5-MP-DOMINO uses implicit aerosol corrections by assuming aerosols to be "effective clouds", as assumed in most satellite NO₂ products except POMINO (Liu et al., 2019; Lin et al., 2014, 2015). In addition, TM5-MP-DOMINO employs a priori NO₂ profiles from the TM5 model at a relatively coarse horizontal resolution ($1^{\circ} \times 1^{\circ}$; Williams et al., 2017). The implicit aerosol corrections (Lin et al., 2014; Lorente et al., 2017; Liu et al., 2019) and the coarse horizontal resolution of a priori NO₂ profile data (Laughner et al., 2016; Russell et al., 2011; McLinden et al., 2014) may be the largest sources of the large biases observed between TM5-MP-DOMINO and MAX-DOAS. Based on previous studies for OMI, implicit aerosol corrections can lead to uncertainties of more than 50 % over polluted areas with high aerosol loadings like China (Lin et al., 2014; Lorente et al., 2017; Liu et al., 2019). Eskes et al. (2018) showed that using a priori NO₂ profiles from the regional CAMS model at a 12 km × 12 km resolution to replace the TM5 NO₂ profiles increases the retrieved NO₂ VCDs by $\sim 0\%$ to 50% over western Europe depending on the location.

Here we present a new TROPOMI tropospheric NO₂ VCD product over East Asia, namely POMINO-TROPOMI. This product is based on our POMINO algorithm (Liu et al., 2019; Lin et al., 2014, 2015) previously applied to OMI. POMINO-TROPOMI improves upon TM5-MP-DOMINO by employing explicit aerosol corrections and using high-resolution (~ 25 km) a priori NO₂ profiles, among other improvements. POMINO-TROPOMI NO₂ VCD data over July–October 2018 are presented and validated by MAX-DOAS measurements, along with additional sensitivity tests of the effects of aerosol representations and a priori NO₂ profiles.

2 Method and data

2.1 POMINO-TROPOMI retrieval algorithm and product

As one of the UV–Vis backscatter instruments to observe NO₂, TROPOMI inherits much of the design of OMI (Veefkind et al., 2012). Thus the POMINO-TROPOMI algorithm here largely follows our previous POMINO algorithm (Liu et al., 2019), with some modifications to adapt to its high horizontal resolution and different cloud retrieval procedure.

The POMINO-TROPOMI algorithm focuses on improving the calculation of tropospheric AMF. It thus takes the tropospheric SCD data from TM5-MP-DOMINO (OFFLINE), which are obtained by fitting the 405–465 nm wavelength range with the DOAS method. Our tropospheric AMF calculation is done for 437.5 nm, following TM5-MP-DOMINO.

We use the parallelized linearized discrete ordinate radiative transfer (LIDORT)-driven AMFv6 package to derive tropospheric AMFs via online pixel-specific radiative transfer calculations, with no use of lookup tables. Our algorithm explicitly accounts for aerosol optical effects and anisotropic properties of surface reflectance, uses daily a priori aerosol and NO₂ profiles from the simulation of nested GEOS-Chem model (0.25° lat $\times 0.3125^{\circ}$ long; Zhang et al., 2016), and further uses aerosol optical depth (AOD) data from Moderate Resolution Imaging Spectroradiometer (MODIS/Aqua) to correct GEOS-Chem simulated aerosols on a monthly basis. Figure 1 shows the procedure of using the AMFv6 package to derive the tropospheric NO₂ VCDs of POMINO-TROPOMI. Table S1 in the Supplement shows the retrieval parameters in POMINO-TROPOMI, in comparison with those in TM5-MP-DOMINO and POMINO v2.

The independent pixel approximation (IPA) is used to calculate AMF as a linear combination of a cloudy AMF (M_{cld})



Figure 1. Flowchart of the POMINO-TROPOMI algorithm. The grey rectangles represent the parameters from the TM5-MP-DOMINO (OFFLINE) product.

and a clear-sky AMF (M_{clr}) as follows:

$$M = wM_{\rm cld} + (1 - w)M_{\rm clr}.$$
(1)

w is the cloud radiation fraction (CRF) calculated by

$$w = \frac{f_{\text{eff}}I_{\text{cld}}}{R} = \frac{f_{\text{eff}}I_{\text{cld}}}{f_{\text{eff}}I_{\text{cld}} + (1 - f_{\text{eff}})I_{\text{clr}}},$$
(2)

where I_{cld} denotes the radiance from the cloudy part of the pixel, I_{clr} the radiance from the clear-sky part of the pixel, f_{eff} the cloud fraction (CF), and *R* the total scene radiance. Retrieval of cloud properties is a prerequisite for NO₂ retrieval. We take the effective cloud pressure (CP) from the FRESCO-S algorithm (van Geffen et al., 2019), which uses the O₂ A band (around 758 nm) for TROPOMI trace gas retrievals. We recalculate *w* and f_{eff} at the NO₂ fitting wavelength (437.5 nm). The online CF calculation is similar to that for TM5-MP-DOMINO (Arnoud et al., 2017; van Geffen et al., 2019) but with an explicit aerosol correction to be consistent with the following NO₂ retrieval and to remove the aerosol signal from the retrieved CF data.

For explicit aerosol corrections in this study, we take daily aerosol simulation results from the GEOS-Chem v9-02 nested model over East Asia, followed by a monthly AOD correction using MODIS/Aqua C6.1 AOD data. Our future study will use the Cloud-Aerosol Lidar with Orthogonal Polarization (CALIOP) aerosol extinction vertical profiles to further improve the modeled aerosol profiles. Figure 2b shows the spatial distribution of AOD in July 2018 used in clouds and NO₂ retrievals. The AOD distribution is consistent (R = 0.42; N = 1447) with that of near-surface PM_{2.5} mass concentration measurements (Fig. 2a) taken from the Ministry of Ecology and Environment of China (MEE); the difference between AOD and near-surface PM_{2.5} is expected because they represent different parameters of aerosols.



Figure 2. (a) Observed near-surface $PM_{2.5}$ mass concentrations averaged over July 2018. Results are sampled at the times of valid TROPOMI data. (b) AOD data on a $0.05^{\circ} \times 0.05^{\circ}$ grid used for the retrieval of POMINO-TROPOMI NO₂ VCDs in July 2018.

The criteria to select valid pixels in this study are as follows. We exclude pixels with viewing zenith angles (VZAs) greater than 80°, with high albedos caused by ice or snow on the ground, or with a quality flag (from TM5-MP-DOMINO) of less than 0.5. To screen out cloudy scenes, we discard pixels with CRFs greater than 50 % in the POMINO-TROPOMI product.

In addition to our formal POMINO-TROPOMI product (referred to as Case REF), we use sensitivity cases to evaluate the impacts of aerosol corrections (explicit versus implicit) and the horizontal resolution of a priori NO₂ profiles (Cases 1 and 2 in Table 1). Two additional cases (Cases 3 and 4) concern the treatment of CP in combination with the choice of aerosols and surface reflectance. Specifically, using the CP data directly from FRESCO-S means that our retrieval algorithm does not perfectly account for the effect of aerosols on clouds. Our retrievals consider the anisotropy of the surface via (bidirectional reflectance distribution function) BRDF effects, while a Lambertian surface is used in

| ID | A priori NO ₂ profiles | Aerosols | Surface reflectance |
|---------------------------|---|----------------------------|----------------------|
| Case REF (POMINO-TROPOMI) | 0.25° lat $\times 0.3125^{\circ}$ long | Explicit | MODIS BRDF |
| Case 1 | Same as Case REF | n/a | Same as Case REF |
| Case 2 | 2° lat $\times 2.5^{\circ}$ long | n/a | Same as Case REF |
| Case 3 | Same as Case REF | Semi-explicit ¹ | Same as Case REF |
| Case 4 | Same as Case REF | Same as Case REF | OMI LER ² |

Table 1. Sensitivity experiments for the NO₂ retrieval based on the POMINO-TROPOMI algorithm (n/a: not applicable).

¹ Explicit aerosol treatments for M_{clr} and no aerosol corrections for M_{cld} . ² The LER dataset is a 5-year climatology built upon OMI measurements on a grid of $0.5^{\circ} \times 0.5^{\circ}$. The dataset is taken from the TM5-MP-DOMINO product.



Figure 3. Spatial distributions of POMINO-TROPOMI NO₂ VCDs (on a $0.05^{\circ} \times 0.05^{\circ}$ grid) around (a) Nanjing and (b) Xuzhou MAX-DOAS measurement sites in July 2018. The MAX-DOAS sites are marked as "+".

deriving the FRESCO-S CP. To ensure sampling consistency, the pixels used in all cases are selected based on the CRF values in Case REF.

2.2 Ground-based MAX-DOAS measurements

We use ground-based MAX-DOAS measurements to validate the POMINO-TROPOMI NO₂ product. The MAX-DOAS measurements are from two suburban stations (Xuzhou and Nanjing) and one remote station (Fukue; Kanaya et al., 2014). Table 2 shows the geographical and time information of each MAX-DOAS site, and Sect. S1 in the Supplement describes each instrument in detail. Although Xuzhou and Nanjing are both classified as suburban sites located at university campuses, the NO₂ spatial distributions around the two sites are very different. The spatial distribution of NO₂ VCDs is relatively smooth around Xuzhou, whereas the VCDs exhibit a strong spatial gradient around Nanjing (Fig. 3).

A consistent spatiotemporal sampling is crucial in comparing satellite measurements and MAX-DOAS data (Boersma et al., 2018; Lin et al., 2014; Liu et al., 2019; Wang et al., 2017). We average all valid MAX-DOAS data within ± 1 h of the TROPOMI overpass time to obtain daily values for comparison. To reduce the influence of local events, we exclude all MAX-DOAS data whose standard deviations within the 2 h exceed 20 % of their mean values. We average all valid pixels within 5 km of each MAX-DOAS site to represent the respective daily satellite data. Section S2 shows how the validation results are affected by the sampling choice.

3 Results

3.1 POMINO-TROPOMI NO₂ VCDs over East Asia

Figure 4b shows the spatial distribution of POMINO-TROPOMI tropospheric NO₂ VCDs over East Asia on a $0.05^{\circ} \times 0.05^{\circ}$ grid in July 2018. High VCD values (> 3×10^{15} molec. cm⁻²) are shown over polluted areas such as East China and India, and low values (< 1×10^{15} molec. cm⁻²) are shown over the open ocean and the Tibetan Plateau. For comparison, the colored dots in Fig. 4a visualize the near-surface NO₂ concentrations observed at the MEE sites at the overpass time of TROPOMI. In both the VCD and the near-surface concentration maps (Fig. 4a and b), hotspots like urban centers and isolated sources can be seen clearly, due to the short lifetimes of NO_x in summer. The spatial correlation is about 0.55 (N = 1458) between the VCD and the near-surface concentration distributions.

Figure 4c shows the spatial distribution of TM5-MP-DOMINO (OFFLINE) NO₂ VCDs for comparison. The general distribution of TM5-MP-DOMINO is consistent with that of POMINO-TROPOMI with a correlation coefficient of 0.97 (N = 1091154). However, TM5-MP-DOMINO NO₂ VCD values are lower than POMINO-TROPOMI by about 35% averaged over the whole domain (Fig. 4d), by -37%-68% over cleaner areas (POMINO-TROPOMI $< 5 \times 10^{15}$ molec. cm⁻²), and by 0%-66% over more polluted areas (POMINO-TROPOMI $\ge 5 \times 10^{15}$ molec. cm⁻²). TM5-MP-DOMINO does not show strong local signals at pollution hotspots like the urban center of Beijing (Fig. 4c). Over these hotspot locations, TM5-MP-DOMINO is lower than POMINO-TROPOMI by up to 5×10^{15} molec. cm⁻² (Fig. 4e).

Figure 5a and b present the two products over Beijing and surrounding areas, showing a much weaker spatial gradient of NO_2 VCDs from Beijing urban center to its outskirts in

Table 2. MAX-DOAS measurement sites.

| Site name | Geographical location | Measurement time |
|-----------|-----------------------------|--------------------------|
| Nanjing | 118.950° E, 32.118° N, 36 m | 1 July-31 October 2018 |
| Xuzhou | 117.142° E, 34.217° N, 92 m | 1 July-31 October 2018 |
| Fukue | 128.680° E, 32.750° N, 83 m | 1 July-15 September 2018 |



Figure 4. The spatial distribution of (**a**) near-surface NO₂ concentrations at monitoring stations, (**b**) POMINO-TROPOMI NO₂ VCD, and (**c**) TM5-MP-DOMINO (OFFLINE) NO₂ VCD at $0.05^{\circ} \times 0.05^{\circ}$ grid in July 2018. Panels (**d**) and (**e**) are relative and absolute difference of TM5-MP-DOMINO (OFFLINE) to POMINO-TROPOMI NO₂ VCD.

TM5-MP-DOMINO than in POMINO-TROPOMI. The corresponding histograms and Gaussian fittings in Fig. 5c also show a lower mean value and a smaller standard deviation of TM5-MP-DOMINO than POMINO-TROPOMI. These results highlight the important differences between the two products at fine scales.

We further compare the two satellite products with ground-based MAX-DOAS measurements at three sites. The scatterplots in Fig. 6a and b compare the NO₂ VCDs on 63 d (from 109 pixels) over July–October 2018 with their MAX-DOAS counterparts. Different colors differentiate the sites. POMINO-TROPOMI captures the day-to-day variability in MAX-DOAS data ($R^2 = 0.75$) and shows a small normalized mean bias (NMB = 0.8 %). The reduced major axis (RMA) regression shows a slope of 0.70, mainly because of the underestimate on high-NO₂ days. TM5-MP-DOMINO is also correlated with MAX-DOAS ($R^2 = 0.68$), although the correlation is weaker than our retrieval. The NMB of TM5-MP-DOMINO is much more significant (-41.9 %), and the

RMA regression slope is much smaller (0.42). Similar underestimates of TM5-MP-DOMINO have been discussed in their "Readme" document (Eskes et al., 2019) and the algorithm theoretical basis document (ATBD) file (van Geffen et al., 2019) in general, in Griffin et al. (2019) for Canada. Major plausible causes of the underestimate of TM5-MP-DOMINO include coarse-resolution climatological surface albedo data, coarse-resolution ($1^{\circ} \times 1^{\circ}$) a priori NO₂ profiles, implicit aerosol corrections, and uncertainties in CP from FRESCO-S.

Cloud pressures from FRESCO-S are found to be too high – i.e., the cloud top is too close to the ground, especially over China (van Geffen et al., 2019). We examine this effect by excluding the pixels with CP > 850 hPa when comparing with MAX-DOAS data. With this additional criterion, the number of valid days drops to 20. Figure 6c and d show the scatterplots and corresponding RMA results. The NMB of TM5-MP-DOMINO is reduced slightly to -38.2%, and its R^2 for day-to-day variation



Figure 5. Spatial distributions of (a) POMINO-TROPOMI NO₂ VCDs and (b) TM5-MP-DOMINO (OFFLINE) NO₂ VCDs on a $0.05^{\circ} \times 0.05^{\circ}$ grid over Beijing and its surrounding areas averaged over July 2018. (c) Histograms of monthly NO₂ VCDs over this region. The bin size is 0.2×10^{15} molec. cm⁻². The black and orange dashed lines are corresponding Gaussian curve fitting of the histograms. μ is the mean value, and σ is the standard deviation of a Gaussian curve fitting.

is increased from 0.68 to 0.85. POMINO-TROPOMI still outperforms TM5-MP-DOMINO: 0.85 versus 0.85 for R^2 , 13.8% versus -38.2% for NMB, and 0.82 versus 0.56 for RMA regression slope. The improvement from excluding CP > 850 hPa scenes is larger in TM5-MP-DOMINO (with implicit aerosol corrections) than in POMINO-TROPOMI (with explicit aerosol corrections). The averaged CF of data excluding CP > 850 hPa is 0.13 (AOD = 0.63), which is much larger than the averaged value (CF = 0.06) in Fig. 6a and b (AOD = 0.57). This appears to imply that the overestimated CP may be partly because the FRESCO-S cloud algorithm might misinterpret heavy aerosol loadings near the ground as clouds, a common issue in satellite data (Lin and Li, 2016).

3.2 Influences of aerosol correction approaches and horizontal resolutions of a priori NO₂ profiles

To investigate the causes of difference between POMINO-TROPOMI and TM5-MP-DOMINO (OFFLINE), we conduct two sensitivity retrievals based on the POMINO-TROPOMI algorithm (Cases 1 and 2 in Table 1). Figure 7 shows the relative (panels a–c) and absolute (panels d–f) dif-

ferences between retrieval cases (REF, Case 1, and Case 2). Case 1 adopts the implicit aerosol correction for both clouds and NO₂ retrievals, as in TM5-MP-DOMINO, so the difference between REF and Case 1 indicates the effect of aerosol representation (explicit versus implicit) (Fig. 7a, d). In Case 2, the high-resolution $(0.25^{\circ} \text{ lat} \times 0.3125^{\circ} \text{ long})$ NO₂ profiles are replaced by low-resolution (2° lat \times 2.5° long) profiles simulated by the GEOS-Chem global model; aerosols are represented implicitly as in Case 1. Case 2 thus mimics TM5-MP-DOMINO, which uses an implicit aerosol correction and coarse-resolution NO2 profiles. Thus, the difference between Case 1 and Case 2 arises from the a priori NO₂ profiles (Fig. 7b, e). The difference between Case 2 and REF further indicates the joint effect of using coarseresolution a priori NO₂ profiles and an implicit aerosol correction (Fig. 7c, f).

Figure 7 shows that the individual influences of aerosol representations (explicit versus implicit) and a priori NO₂ profiles (fine versus coarse resolution) vary substantially from one location to another. The impacts of aerosol corrections are most evident over the areas of heavy aerosol loadings including East China, India, and parts of Southeast Asia.



Figure 6. Scatterplot for daily NO₂ VCDs (10^{15} molec. cm⁻²) between MAX-DOAS and two TROPOMI NO₂ VCD products. Each colored dot represents a day, and each color denotes a station. For each day, the satellite data are averaged over all pixels. Panels (c) and (d) are results of two TROPOMI NO₂ VCD products with effective cloud pressures ≤ 850 hPa.

The implicit aerosol correction (Case 1) tends to result in lower NO₂ VCDs by 0%-50% over urban areas compared to the explicit aerosol correction (Case REF) (Fig. 7a, d). By comparison, the impacts of NO₂ profiles are spatially more heterogeneous (Fig. 7b and e). Over the offshore coastal areas, using coarse-resolution NO₂ profiles (Case 2) tends to overestimate the NO₂ VCDs by 30%-100% relative to when high-resolution profiles are used (Case 1), due to (horizontal) over-smoothing of NO₂.

Below, we discuss these differences over two key areas including northern East China and Xinjiang. Northern Eastern China (bounded by the black rectangle in Fig. 7a) is a region with heavy aerosol loading region (Fig. 2). Over this region, an implicit rather than explicit aerosol representation results in lower NO₂ VCDs by $\sim 25 \%$ (Fig. 7a, d), while the effect of NO₂ profiles is weaker (Fig. 7b, e). The joint effect is dominated by the effect of aerosol representation (Fig. 7c, f).

That the impact of a priori NO_2 profiles is relatively small over northern Eastern China is partly because of the smearing effect by wind. Figure 8 differentiates the effects of NO_2 profiles between 23 windy (daily average wind speed under $500 \text{ m} > 2 \text{ m s}^{-1}$) days and 7 calm (wind speed $< 2 \text{ m s}^{-1}$) days. The dataset of wind is taken from the National Aeronautics and Space Administration–Global Modeling and Assimilation Office's (NASA–GMAO's) "forward-processing" (GEOS-FP) data product with the horizontal resolution at 0.25° lat × 0.3125° long. In the windy cases, the NO₂ VCDs are much more smoothed even at 0.3125° resolution, and thus the difference of NO₂ resolutions is very small. In contrast, NO₂ VCDs exhibit much stronger horizontal gradient in calm situations, which are better retrieved when high-resolution NO₂ profiles are used. As calm situation is more favorable for pollutant accumulation, while windy days help to dilute concentration of NO₂, so much higher NO₂ VCDs are found in Fig. 8a and b.

Xinjiang (bounded by the red rectangle in Fig. 7a) is a deserted region in West China. Over Xinjiang, the resolution of a priori NO_2 profiles affects the retrieved NO_2 VCDs much more than the aerosol representation does (Fig. 7b and e). Compared to Case 1 (with high-resolution NO_2 profiles), Case 2 (with coarse-resolution profiles) leads to much lower



Figure 7. (a–c) Relative differences caused by aerosol corrections and a priori NO₂ profiles. Panels (d)–(f) are the corresponding absolute differences. The black and red rectangles stand for northern Eastern China and Xinjiang, respectively.

NO₂ VCDs over the isolated urban areas which are not resolved by the coarse model.

3.3 Influences of directly using the CP data from FRESCO-S

As we take the CP data directly from the FRESCO-S retrieval rather than re-retrieving CP (as done for CF), two main issues arise. First, the FRESCO-S retrieved CP may be affected by aerosols, and thus using such CP data in our explicit aerosol corrections (Case REF) may lead to overcorrection of aerosol effects. To estimate the effect of such overcorrection on retrieved NO₂ VCDs, we employ in an additional sensitivity case (Case 3 in Table 1) a "semi-explicit" aerosol correction approach. This approach explicitly includes aerosols in the calculation of AMF for the clear-sky portion (M_{clr}) of a pixel (as in Case REF) but excludes aerosols for the cloudy-sky portion (M_{cld}) of that pixel. Correspondingly, CF is recalculated on the basis that the radiance at 437.5 nm received by TROPOMI is contributed from the aerosol-contained clearsky portion and the no-aerosol, cloudy-sky portion. Table 3 shows that in July 2018, on a pixel basis, the derived NO₂ profiles in Case 3 are larger than those in Case REF, with an average difference increasing from 3.1 % at relatively clean situations (NO₂ VCDs in Case REF $< 5 \times 10^{15}$ molec. cm⁻²) to 11.2 % for polluted situations (NO₂ VCDs in Case REF \geq 15×10^{15} molec. cm⁻²). The spatial distributions in Fig. S1a and S1b also show higher NO2 VCDs in Case 3 relative to Case REF. The corresponding increases in CF (Fig. S1c versus Fig. S1d) are because in Case 3 the scattering contributions to the radiance from aerosols in the cloudy-sky portion (that would have occurred) are accounted for with higher CFs. The enhanced "shielding effect" of clouds (due to higher CFs) result in lower NO₂ AMFs and higher VCDs.



Figure 8. Spatial distributions of POMINO-TROPOMI NO₂ VCDs on a $0.05^{\circ} \times 0.05^{\circ}$ grid under calm conditions with (a) 0.25° lat $\times 0.3125^{\circ}$ long NO₂ profiles (Case 1), with (b) 2° lat $\times 2.5^{\circ}$ long profiles (Case 2), and (c) their relative differences. Panels (d)–(f) are similar to panels (a)–(c) but under windy conditions.

Table 3. Effects of choices of aerosols and surface reflectance inconsistent with using the CP data from FRESCO-S in July 2018

| Situation ^a | Effect of aerosol correction choice (Case 3 – REF) ^b | Effect of surface reflectance choice (Case 4 – REF) ^b |
|---|---|--|
| NO ₂ VCD $< 5 \times 10^{15}$ molec. cm ⁻² Mean AOD: 0.27 Mean CF: 0.07 Mean CP: 748 hPa | 3.1 % | -3.7% |
| $\begin{array}{l} 5\times10^{15}\ \text{molec.}\ \text{cm}^{-2}\leq\text{NO}_2\ \text{VCD}\\ <15\times10^{15}\ \text{molec.}\ \text{cm}^{-2}\\ \text{Mean AOD:}\ 0.65\\ \text{Mean CF:}\ 0.08\\ \text{Mean CP:}\ 767\ \text{hPa} \end{array}$ | 4.1 % | -8.1% |
| 15×10^{15} molec. cm ⁻² \leq NO ₂ VCD Mean AOD: 0.66 Mean CF: 0.09 Mean CP: 772 hPa | 11.2 % | -8.3 % |

^a The values of AOD, CF, and CP shown here are mean values of the pixels of corresponding subsets in

Case REF. ^b The percentage values represent the mean relative differences relative to Case REF.

For surface reflectance, Case REF considers the BRDF effect instead of Lambertian equivalent reflectivity (LER) which is used in deriving FRESCO-S clouds. This leads to inconsistency between when the CP is derived and when it is used. The LER data used by FRESCO-S are gener-

ated at 758 and 772 nm (based on the Global Ozone Monitoring Experiment-2), rather than at the 437.5 nm for the NO₂ retrieval. Thus Case 4 adopts the OMI LER data from TM5-MP-DOMINO – a 5-year, monthly-based climatology at 440 nm – and recalculates CFs and NO₂ AMFs with ex-

| Error sou | rce | Estimated error in parameter | Corresponding error in tropospheric NO ₂ AMF |
|----------------------|-------------------------|---|--|
| NO ₂ prof | iles | $+0.01^{b}$ | $\pm 10\%^{a}$ $\pm 10\%^{b}$ |
| Cloud pre | essure | $\pm 50 \mathrm{hPa^{c}}$ | $\pm 10\%$ $\pm 10\%^{c}$ |
| BRDF co | efficients | ±10 % ^u | $\pm 10 \%^{\rm u}$ |
| Aerosol | AOD | ±0.07 ^e | $\pm 10\%$ for clean situations |
| | SSA | $\pm 0.03^{\rm f}$ | Up to $\pm 20\%$ for heavily polluted |
| | Aerosol layer height | Underestimated by 300–600 m ^g | cases ^h |

Table 4. Contributions of estimated errors in POMINO-TROPOMI NO2 AMFs.

^a This error estimate is based on Lin et al. (2010), Lin et al. (2014), Boersma et al. (2004, 2011, 2018) and this study. The error accounts for the effect of horizontal resolutions and the vertical process in GEOS-Chem. ^b This accounts for the expectation that our explicit aerosol correction leads to more reasonable CFs. ^c Based on van Geffen et al. (2019). Our estimated NO₂ errors related to the use of FRESCO CP data (instead of recalculating it) are within this error range. ^d The estimated parameter error is taken from Zhou et al. (2010), and the corresponding AMF error is based on Lin et al. (2014) and Case 4. ^e Based on Tian et al. (2015), which compared the MODIS Merged AOD C6.1 product with AERONET in China. ^f Based on Lin et al. (2015), which compared GEOS-Chem simulated aerosol extinction profiles with CALIOP. ^h This is a tentative error estimate based on Lin et al. (2015), Lin et al. (2017), Liu et al. (2019), and this study. With explicit aerosol corrections, the errors in heavily polluted situations are expected to be smaller than when assuming implicit aerosol corrections.

plicit aerosol corrections and FRESCO-S CP (Table 1). Here, the ice/snow flag in the TM5-MP-DOMINO product is used to exclude the possible ice/snow contamination, and only the pixels with blue-sky albedos (derived from the BRDF data in Case REF) less than 0.3 are taken into consideration. The resulting NO₂ VCDs in Case 4 are lower than Case REF by 3.7 % on a pixel-based average for relatively clean situations and by 8.3 % for polluted situations (Table 3). Figure S2a and S2b further show the spatial distributions of the relative and absolute differences in derived monthly mean NO₂ VCDs between Case 4 and Case REF. In general, Case 4 leads to lower NO₂ VCDs than Case REF because of stronger surface reflectance, as is obvious in the comparison of blue-sky albedo in Case REF and LER albedo in Case 4 (Fig. S2c versus Fig. S2d).

3.4 Error estimates for POMINO-TROPOMI NO₂ AMFs

It is difficult to derive an overall AMF error for each pixel with our algorithm, particularly because of the threedimensional aerosol parameters used, the interlinkage between clouds and other parameters (aerosols and surface reflectance), and the exclusion of lookup tables (LUTs; which leads to too computationally expensive error estimates). Table 4 provides a preliminary estimate of the NO₂ AMF errors with respect to uncertainties in individual parameters. We follow the ATBD of TM5-MP-DOMINO (van Geffen et al., 2019) and make use of error estimates in previous studies. Individual errors with respect to CF, CP, and BRDF coefficients are within 10%. Errors with respect to aerosols are considered to be larger, due to uncertainty in AOD, single-scattering albedo (SSA), and aerosol vertical profiles (Liu et al., 2019), as well as the fact that using the CP from FRESCO-S rather than deriving it here may lead to an additional error in the NO₂ AMFs. Note that these individual errors are not fully independent due to the aforementioned linkage between parameters. The magnitude of potential systematic bias in NO₂ may be smaller than the quadrature sum of errors in individual parameters, as suggested by the slight mean bias relative to MAX-DOAS data (Fig. 6a).

4 Conclusion and discussion

The POMINO algorithm to retrieve tropospheric NO₂ VCDs has been successfully applied to TROPOMI over East Asia. The resulting POMINO-TROPOMI product shows higher tropospheric NO₂ VCDs (by about 35% averaged over East Asia) and much clearer urban and other hotspot signals, compared to the TM5-MP-DOMINO (OFFLINE) product. Further evaluation using independent MAX-DOAS measurements indicates very good performance of POMINO-TROPOMI in capturing the day-to-day variation ($R^2 = 0.75$; N = 63) and mean value (NMB = 0.8%) of NO₂, better than TM5-MP-DOMINO (0.68 and -41.9%, respectively).

Over regions with heavy aerosol loading, the accuracy of retrieved NO₂ VCDs is affected substantially by how aerosols are represented in the retrieval process (implicit or explicit). The implicit aerosol representation tends to underestimate the NO₂ VCDs by 0%-50% over most urban areas in East Asia and by about 25% over northern East

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China. Using a priori NO₂ profile data at a horizontal resolution of ~ 25 km, POMINO-TROPOMI captures the cityscale NO₂ hotspots. Reducing the horizontal resolution of a priori profiles to what is typically set up by global models (100–200 km) underestimates the NO₂ hotspots and the spatial gradient surrounding them, and the effects are more pronounced in calm than in windy situations. Overall, our results provide useful information to improve TROPOMI retrieval algorithms and offer insight for applications to the upcoming geostationary satellite instruments including GEMS, TEMPO, and Sentinel-4 Precursor.

Further work can be done to improve the retrieval algorithm for TROPOMI. First, the hybrid cloud retrieval method used in POMINO-TROPOMI is not optimal, because only cloud fraction but not cloud pressure is recalculated with explicit aerosol corrections and BRDF effects. The uncertainty caused by inconsistent assumptions of aerosols and albedos in cloud pressure and NO2 VCD retrievals are initially estimated in this study. Recalculation of cloud pressure will be done in the future using the O_2-O_2 method when the O_2-O_2 O₂ SCD data are available. Second, correcting the simulated aerosol extinction vertical profiles will further improve the clouds and NO₂ retrievals (Liu et al., 2019). Third, the horizontal resolution of a priori NO₂ profiles (~ 25 km) does not match the fine footprint of TROPOMI, and further increasing the resolution will help achieve better accuracy for analyses of very fine scale pollution characteristics such as along the highways and rivers and within urban centers.

Data availability. The POMINO-TROPOMI NO₂ data are available at the ACM group product website (http: //www.pku-atmos-acm.org/acmProduct.php/, Lin et al., 2020). The TM5-MP-DOMINO product can be download via the TEMIS website at http://www.temis.nl/airpollution/no2.html Royal Netherlands Meteorological Institute, 2020). The near-surface data of NO₂ and PM_{2.5} can be downloaded from http://www.cnemc.cn/sssj/cskqzl/ (Ministry of Ecology and Environment, 2019). The ground-based MAX-DOAS observations would be provided after the applications of users are approved by corresponding owners.

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References

- Arnoud, A., Mattia, P., Maarten, S., Veefkind, J. P., Loyola, D., and Wang, P.: Sentinel-5 precursor/TROPOMI Level2 Product User Manual KNMI level2 support products, Royal Netherlands Meteorological Institute (KNMI), De Bilt, the Netherlands, ATBD, 2017.
- Boersma, K. F., Eskes, H. J., and Brinksma, E. J.: Error analysis for tropospheric NO₂ retrieval from space, J. Geophys. Res.-Atmos., 109, D04311, https://doi.org/10.1029/2003JD003962, 2004.
- Boersma, K. F., Eskes, H. J., Dirksen, R. J., van der A, R. J., Veefkind, J. P., Stammes, P., Huijnen, V., Kleipool, Q. L., Sneep, M., Claas, J., Leitão, J., Richter, A., Zhou, Y., and Brunner, D.: An improved tropospheric NO₂ column retrieval algorithm for the Ozone Monitoring Instrument, Atmos. Meas. Tech., 4, 1905– 1928, https://doi.org/10.5194/amt-4-1905-2011, 2011.
- Boersma, K. F., Eskes, H. J., Richter, A., De Smedt, I., Lorente, A., Beirle, S., van Geffen, J. H. G. M., Zara, M., Peters, E., Van Roozendael, M., Wagner, T., Maasakkers, J. D., van der A, R. J., Nightingale, J., De Rudder, A., Irie, H., Pinardi, G., Lambert, J.-C., and Compernolle, S. C.: Improving algorithms and uncertainty estimates for satellite NO₂ retrievals: results from the quality assurance for the essential climate variables (QA4ECV) project, Atmos. Meas. Tech., 11, 6651–6678, https://doi.org/10.5194/amt-11-6651-2018, 2018.
- Dentener, F., van Weele, M., Krol, M., Houweling, S., and van Velthoven, P.: Trends and inter-annual variability of methane emissions derived from 1979-1993 global CTM simulations, Atmos. Chem. Phys., 3, 73–88, https://doi.org/10.5194/acp-3-73-2003, 2003.
- Eskes, H., van Geffen, J. H. G. M., Boersma, F., Sneep, M., ter Linden, M., and Veefkind, P.: Sentinel-5P TROPOMI highresolution nitrogen dioxide observations, American Geophysical Union Fall Meeting, 14 December 2018, Walter E Washington Convention Center, Washington, DC, USA, A51A-02, 2018.
- Eskes, H. J., Eichmann, K. U., Lambert, J. C., Loyola, D., Veefkind, J. P., Dehn, A., and Zehner, C.: S5P Mission Performance Centre Nitrogen Dioxide [L2_NO2_] Readme, Royal Netherlands

Meteorological Institute (KNMI) De Bilt, the Netherlands, version 01.03.00, 2019.

- Goryl, P., Zehner, C., and Laur, H.: Sentinel-5 Precursor Calibration and Validation Plan for the Operational Phase, Royal Netherlands Meteorological Institute (KNMI), De Bilt, the Netherlands, 2017.
- Griffin, D., Zhao, X., McLinden, C. A., Boersma, F., Bourassa, A., Dammers, E., Degenstein, D., Eskes, H., Fehr, L., Fioletov, V., Hayden, K., Kharol, S. K., Li, S.-M., Makar, P., Martin, R. V., Mihele, C., Mittermeier, R. L., Krotkov, N., Sneep, M., Lamsal, L. N., Linden, M. t., Geffen, J. v., Veefkind, P., and Wolde, M.: High-Resolution Mapping of Nitrogen Dioxide With TROPOMI: First Results and Validation Over the Canadian Oil Sands, Geophys. Res. Lett., 46, 1049–1060, https://doi.org/10.1029/2018g1081095, 2019.
- Hoek, G., Krishnan, R. M., Beelen, R., Peters, A., Ostro, B., Brunekreef, B., and Kaufman, J. D.: Long-term air pollution exposure and cardio- respiratory mortality: a review, Environ. Health, 12, 43, https://doi.org/10.1186/1476-069x-12-43, 2013.
- Kanaya, Y., Irie, H., Takashima, H., Iwabuchi, H., Akimoto, H., Sudo, K., Gu, M., Chong, J., Kim, Y. J., Lee, H., Li, A., Si, F., Xu, J., Xie, P.-H., Liu, W.-Q., Dzhola, A., Postylyakov, O., Ivanov, V., Grechko, E., Terpugova, S., and Panchenko, M.: Long-term MAX-DOAS network observations of NO₂ in Russia and Asia (MADRAS) during the period 2007–2012: instrumentation, elucidation of climatology, and comparisons with OMI satellite observations and global model simulations, Atmos. Chem. Phys., 14, 7909–7927, https://doi.org/10.5194/acp-14-7909-2014, 2014.
- Krotkov, N. A., McLinden, C. A., Li, C., Lamsal, L. N., Celarier, E. A., Marchenko, S. V., Swartz, W. H., Bucsela, E. J., Joiner, J., Duncan, B. N., Boersma, K. F., Veefkind, J. P., Levelt, P. F., Fioletov, V. E., Dickerson, R. R., He, H., Lu, Z., and Streets, D. G.: Aura OMI observations of regional SO₂ and NO₂ pollution changes from 2005 to 2015, Atmos. Chem. Phys., 16, 4605– 4629, https://doi.org/10.5194/acp-16-4605-2016, 2016.
- Laughner, J. L., Zare, A., and Cohen, R. C.: Effects of daily meteorology on the interpretation of space-based remote sensing of NO₂, Atmos. Chem. Phys., 16, 15247–15264, https://doi.org/10.5194/acp-16-15247-2016, 2016.
- Lee, K. H., Li, Z., Wong, M. S., Xin, J., Wang, Y., Hao, W. M., and Zhao, F.: Aerosol single scattering albedo estimated across China from a combination of ground and satellite measurements, J. Geophys. Res., 112, D22S15, https://doi.org/10.1029/2007JD009077, 2007.
- Lin, J. and Li, J.: Spatio-temporal variability of aerosols over East China inferred by merged visibility-GEOS-Chem aerosol optical depth, Atmos. Environ., 132, 111–122, https://doi.org/10.1016/j.atmosenv.2016.02.037, 2016.
- Lin, J.-T., McElroy, M. B., and Boersma, K. F.: Constraint of anthropogenic NO_x emissions in China from different sectors: a new methodology using multiple satellite retrievals, Atmos. Chem. Phys., 10, 63–78, https://doi.org/10.5194/acp-10-63-2010, 2010.
- Lin, J.-T., Martin, R. V., Boersma, K. F., Sneep, M., Stammes, P., Spurr, R., Wang, P., Van Roozendael, M., Clémer, K., and Irie, H.: Retrieving tropospheric nitrogen dioxide from the Ozone Monitoring Instrument: effects of aerosols, surface reflectance anisotropy, and vertical profile of nitrogen dioxide, At-

mos. Chem. Phys., 14, 1441–1461, https://doi.org/10.5194/acp-14-1441-2014, 2014.

- Lin, J.-T., Liu, M.-Y., Xin, J.-Y., Boersma, K. F., Spurr, R., Martin, R., and Zhang, Q.: Influence of aerosols and surface reflectance on satellite NO₂ retrieval: seasonal and spatial characteristics and implications for NO_x emission constraints, Atmos. Chem. Phys., 15, 11217–11241, https://doi.org/10.5194/acp-15-11217-2015, 2015.
- Lin, J.-T., Liu, M.-Y., Kong, H., and Weng, H.-J.: Peking University ACM group product, available at: http://www.pku-atmos-acm. org/acmProduct.php/, last access: 4 August 2020.
- Liu, M., Lin, J., Boersma, K. F., Pinardi, G., Wang, Y., Chimot, J., Wagner, T., Xie, P., Eskes, H., Van Roozendael, M., Hendrick, F., Wang, P., Wang, T., Yan, Y., Chen, L., and Ni, R.: Improved aerosol correction for OMI tropospheric NO₂ retrieval over East Asia: constraint from CALIOP aerosol vertical profile, Atmos. Meas. Tech., 12, 1–21, https://doi.org/10.5194/amt-12-1-2019, 2019.
- Lorente, A., Folkert Boersma, K., Yu, H., Dörner, S., Hilboll, A., Richter, A., Liu, M., Lamsal, L. N., Barkley, M., De Smedt, I., Van Roozendael, M., Wang, Y., Wagner, T., Beirle, S., Lin, J.-T., Krotkov, N., Stammes, P., Wang, P., Eskes, H. J., and Krol, M.: Structural uncertainty in air mass factor calculation for NO₂ and HCHO satellite retrievals, Atmos. Meas. Tech., 10, 759–782, https://doi.org/10.5194/amt-10-759-2017, 2017.
- McLinden, C. A., Fioletov, V., Boersma, K. F., Kharol, S. K., Krotkov, N., Lamsal, L., Makar, P. A., Martin, R. V., Veefkind, J. P., and Yang, K.: Improved satellite retrievals of NO₂ and SO₂ over the Canadian oil sands and comparisons with surface measurements, Atmos. Chem. Phys., 14, 3637–3656, https://doi.org/10.5194/acp-14-3637-2014, 2014.
- Ministry of Ecology and Environment: China National Environmental Monitoring Centre, available at: http://www.cnemc.cn/ sssj/cskqzl/, last access: 20 December 2019.
- Ott, L. E., Pickering, K. E., Stenchikov, G. L., Allen, D. J., De-Caria, A. J., Ridley, B., Lin, R.-F., Lang, S., and Tao, W.-K.: Production of lightning NO_x and its vertical distribution calculated from three-dimensional cloud-scale chemical transport model simulations, J. Geophys. Res.-Atmos., 115, D04301, https://doi.org/10.1029/2009JD011880, 2010.
- Royal Netherlands Meteorological Institute: Tropospheric Emission Monitoring Internet Service (TEMIS), available at: http://www. temis.nl/airpollution/no2.html, last access: 4 August 2020.
- Russell, A. R., Perring, A. E., Valin, L. C., Bucsela, E. J., Browne, E. C., Wooldridge, P. J., and Cohen, R. C.: A high spatial resolution retrieval of NO₂ column densities from OMI: method and evaluation, Atmos. Chem. Phys., 11, 8543–8554, https://doi.org/10.5194/acp-11-8543-2011, 2011.
- Shindell, D. T., Faluvegi, G., Koch, D. M., Schmidt, G. A., Unger, N., and Bauer, S. E.: Improved Attribution of Climate Forcing to Emissions, Science, 326, 716–718, https://doi.org/10.1126/science.1174760, 2009.
- Tian, X., Xie, P.-H., Xu, J., Li, A., Wang, Y., Qin, M., Hu, Z.-K.: Long-term observations of tropospheric NO₂, SO₂ and HCHO by MAX-DOAS in Yangtze River Delta area, China, J. Environ. Sci., 71, 207–221, https://doi.org/10.1016/j.jes.2018.03.006, 2018.
- van Geffen, J. H. G. M., Eskes, H. J., Boersma, K. F., Maasakkers, J. D., and Veefkind, J. P.: TROPOMI ATBD of the total and tropo-

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spheric NO_2 data products (issue 1.4.0), Royal Netherlands Meteorological Institute (KNMI), De Bilt, the Netherlands, 2019.

- Veefkind, J. P., Aben, I., McMullan, K., Förster, H., de Vries, J., Otter, G., Claas, J., Eskes, H. J., de Haan, J. F., Kleipool, Q., van Weele, M., Hasekamp, O., Hoogeveen, R., Landgraf, J., Snel, R., Tol, P., Ingmann, P., Voors, R., Kruizinga, B., Vink, R., Visser, H., and Levelt, P. F.: TROPOMI on the ESA Sentinel-5 Precursor: A GMES mission for global observations of the atmospheric composition for climate, air quality and ozone layer applications, Remote Sens. Environ., 120, 70–83, https://doi.org/10.1016/j.rse.2011.09.027, 2012.
- Wang, Y., Lampel, J., Xie, P., Beirle, S., Li, A., Wu, D., and Wagner, T.: Ground-based MAX-DOAS observations of tropospheric aerosols, NO₂, SO₂ and HCHO in Wuxi, China, from 2011 to 2014, Atmos. Chem. Phys., 17, 2189–2215, https://doi.org/10.5194/acp-17-2189-2017, 2017.
- Williams, J. E., Boersma, K. F., Le Sager, P., and Verstraeten, W. W.: The high-resolution version of TM5-MP for optimized satellite retrievals: description and validation, Geosci. Model Dev., 10, 721–750, https://doi.org/10.5194/gmd-10-721-2017, 2017.

- Zara, M., Boersma, K. F., De Smedt, I., Richter, A., Peters, E., van Geffen, J. H. G. M., Beirle, S., Wagner, T., Van Roozendael, M., Marchenko, S., Lamsal, L. N., and Eskes, H. J.: Improved slant column density retrieval of nitrogen dioxide and formaldehyde for OMI and GOME-2A from QA4ECV: intercomparison, uncertainty characterisation, and trends, Atmos. Meas. Tech., 11, 4033–4058, https://doi.org/10.5194/amt-11-4033-2018, 2018.
- Zhang, L., Shao, J., Lu, X., Zhao, Y., Hu, Y., Henze, D. K., Liao, H., Gong, S., and Zhang, Q.: Sources and Processes Affecting Fine Particulate Matter Pollution over North China: An Adjoint Analysis of the Beijing APEC Period, Environ. Sci. Technol., 50, 8731–8740, https://doi.org/10.1021/acs.est.6b03010, 2016.
- Zhou, Y., Brunner, D., Spurr, R. J. D., Boersma, K. F., Sneep, M., Popp, C., and Buchmann, B.: Accounting for surface reflectance anisotropy in satellite retrievals of tropospheric NO₂, Atmos. Meas. Tech., 3, 1185–1203, https://doi.org/10.5194/amt-3-1185-2010, 2010.