

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

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26 # Now at Royal Netherlands Meteorological Institute (KNMI)

27 **Abstract**

28 We present a new product with explicit aerosol corrections, POMINO-TROPOMI, for
29 tropospheric nitrogen dioxide (NO₂) vertical column densities (VCDs) over East Asia,
30 based on the newly launched TROPOspheric Monitoring Instrument with an
31 unprecedented high horizontal resolution. Compared to the official
32 TM5-MP-DOMINO (OFFLINE) product, POMINO-TROPOMI shows stronger
33 concentration gradients near emission source locations and better agrees with
34 MAX-DOAS measurements ($R^2 = 0.75$, NMB = 0.8% versus $R^2 = 0.68$, NMB =

35 – 41.9%). Sensitivity tests suggest that implicit aerosol corrections, as in
36 TM5-MP-DOMINO, lead to underestimations of NO₂ columns by about 25% over the
37 polluted Northern East China region. Reducing the horizontal resolution of a priori
38 NO₂ profiles would underestimate the retrieved NO₂ columns over isolated city clusters
39 in western China by 35% but with overestimates by more than 50% over many offshore
40 coastal areas. The effect of a priori NO₂ profiles is more important under calm
41 conditions.

42 **1 Introduction**

43 Nitrogen oxides (NO_x = NO + NO₂) are crucial gaseous pollutants in the troposphere.
44 NO_x lead to the production of particulate matter and ozone (O₃) and enhance levels of
45 oxidants in the troposphere (Shindell et al., 2009), which affect air quality (Dentener
46 et al., 2003) and human health (Hoek et al., 2013). Satellite remote sensing is widely
47 used to monitor levels of nitrogen dioxide (NO₂) pollution worldwide (Krotkov et al.,
48 2016; Lin et al., 2010; McLinden et al., 2014; Ott et al., 2010; Russell et al., 2011).
49 The TROPOspheric Monitoring Instrument (TROPOMI), which was jointly
50 developed by the Netherlands and Europe Space Agency (ESA) (Veefkind et al., 2012)
51 and was launched on October 13th 2017, is a UV/Visible/Near-Infrared/Short-wave
52 infrared backscattering sensor onboard the sun-synchronous Sentinel-5 Precursor
53 (S5P) satellite with an overpass time of 13:30 local solar time. With a wide swath of
54 2,600 km and an unprecedentedly high horizontal resolution of $3.5 \times 7 \text{ km}^2$, ($3.5 \times$
55 5.5 km^2 since 6 August 2019) TROPOMI achieves daily global coverage. This high
56 horizontal resolution and good spatial coverage, combined with the very high
57 signal-to-noise ratio, enables the instrument to resolve NO₂ pollution from point
58 sources, medium-size urban centers, highways or rivers, tasks that were difficult to
59 achieve before.

60 Retrievals of tropospheric NO₂ vertical column densities (VCDs) in the UV/Vis
61 spectral range from satellite instruments consist of three steps: (1) using the
62 Differential Optical Absorption Spectroscopy (DOAS) to fit the slant column density
63 (SCD) of NO₂ along the light path, (2) subtracting the stratospheric contribution from
64 the SCD to obtain the tropospheric SCD, and (3) converting the tropospheric SCD to
65 the tropospheric VCD by using the calculated air mass factor (AMF). For TROPOMI,
66 the random uncertainty of the total SCDs is $\sim 0.6 \times 10^{15} \text{ molec}\cdot\text{cm}^{-2}$, considerably
67 smaller than for the Ozone Monitoring Instrument (OMI, $\sim 0.8 \times 10^{15} \text{ molec}\cdot\text{cm}^{-2}$,
68 (Zara et al., 2018)). The (total or stratospheric) bias is generally between 0 and -10%
69 according to SAOZ observations (Eskes et al., 2019), which meets the error
70 requirements as defined in the S5P Calibration and Validation Plan (Goryl et al.,
71 2017). The calculation of the AMF introduces the dominant source of error in the
72 retrieved tropospheric NO₂ VCDs over polluted areas (Boersma et al., 2004; Boersma
73 et al., 2011; Boersma et al., 2018; Lin et al., 2014; Lorente et al., 2017). The median

74 negative biases of the daily comparisons between tropospheric VCDs from the Dutch
75 official TM5-MP-DOMINO (OFFLINE) product and MAX-DOAS measurements are
76 generally less than 50% (the allowable bias is 25-50% (Goryl et al., 2017)) but quite
77 variable with the stations and NO₂ levels, especially at polluted locations (Eskes et al.,
78 2018; van Geffen, Eskes, Boersma, Maasackers et al., 2019).

79 TM5-MP-DOMINO uses implicit aerosol corrections by assuming aerosols to be
80 “effective clouds”, as assumed in most satellite NO₂ products except POMINO (J. T.
81 Lin et al., 2015; J. T. Lin et al., 2014; Liu et al., 2019). In addition,
82 TM5-MP-DOMINO employs a priori NO₂ profiles from the TM5 model at a
83 relatively coarse horizontal resolution (1° × 1°) (Williams, Boersma, Le Sager, &
84 Verstraeten, 2017). The implicit aerosol corrections (Lin et al., 2014; Liu et al., 2019;
85 Lorente et al., 2017) and the coarse horizontal resolution of a priori NO₂ profile data
86 (Laughner, Zare, & Cohen, 2016; McLinden et al., 2014; Russell et al., 2011) may be
87 the largest sources of the large biases observed between TM5-MP-DOMINO and
88 MAX-DOAS. Based on previous studies for OMI, implicit aerosol corrections can
89 lead to more than 50% uncertainties over polluted areas with high aerosol loadings
90 like China (Lin et al., 2014; Liu et al., 2019; Lorente et al., 2017). Eskes et al. (2018)
91 showed that using a priori NO₂ profiles from the regional CAMS model at a 12 × 12
92 km² resolution to replace the TM5 NO₂ profiles increase the retrieved NO₂ VCDs by
93 ~0 to 50% over Western Europe depending on the location.

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

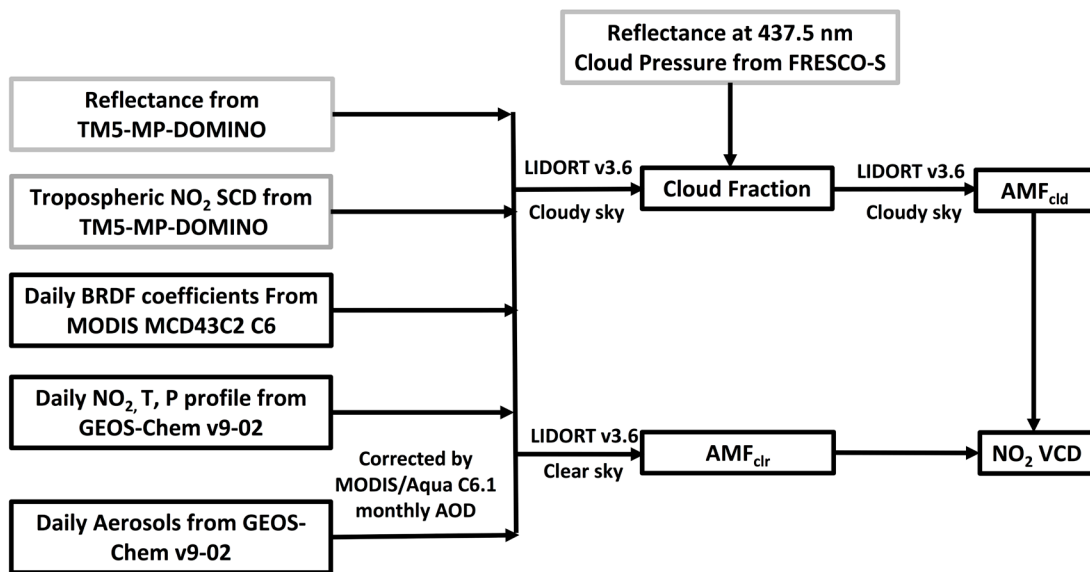
102 **2 Method and Data**

103 *2.1 POMINO-TROPOMI retrieval algorithm and product*

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

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

114 We use the parallelized LIDORT-driven AMFv6 package to derive tropospheric
 115 AMFs via online pixel-specific radiative transfer calculations, with no use of look-up
 116 tables. Our algorithm explicitly accounts for aerosol optical effects and anisotropic
 117 properties of surface reflectance, uses daily a priori aerosol and NO₂ profiles from the
 118 simulation of nested GEOS-Chem model (0.25° lat. × 0.3125° long.), and further
 119 uses Aerosol Optical Depth (AOD) data from Moderate Resolution Imaging
 120 Spectroradiometer (MODIS/Aqua) to correct GEOS-Chem simulated aerosols on a
 121 monthly basis. Figure 1 shows the procedure of using the AMFv6 package to derive
 122 the tropospheric NO₂ VCDs of POMINO-TROPOMI. Table S1 shows the retrieval
 123 parameters in POMINO-TROPOMI, in comparison with those in TM5-MP-DOMINO
 124 and POMINO v2.



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

127

128 The independent pixel approximation (IPA) is used to calculate AMF as a linear
129 combination of a cloudy AMF (M_{cld}) and a clear-sky AMF (M_{clr}):

$$130 \quad M = wM_{\text{cld}} + (1 - w)M_{\text{clr}} \quad (1)$$

131 w is the cloud radiation fraction (CRF) calculated by:

$$132 \quad w = \frac{f_{\text{eff}}I_{\text{cld}}}{R} = \frac{f_{\text{eff}}I_{\text{cld}}}{f_{\text{eff}}I_{\text{cld}} + (1 - f_{\text{eff}})I_{\text{clr}}} \quad (2)$$

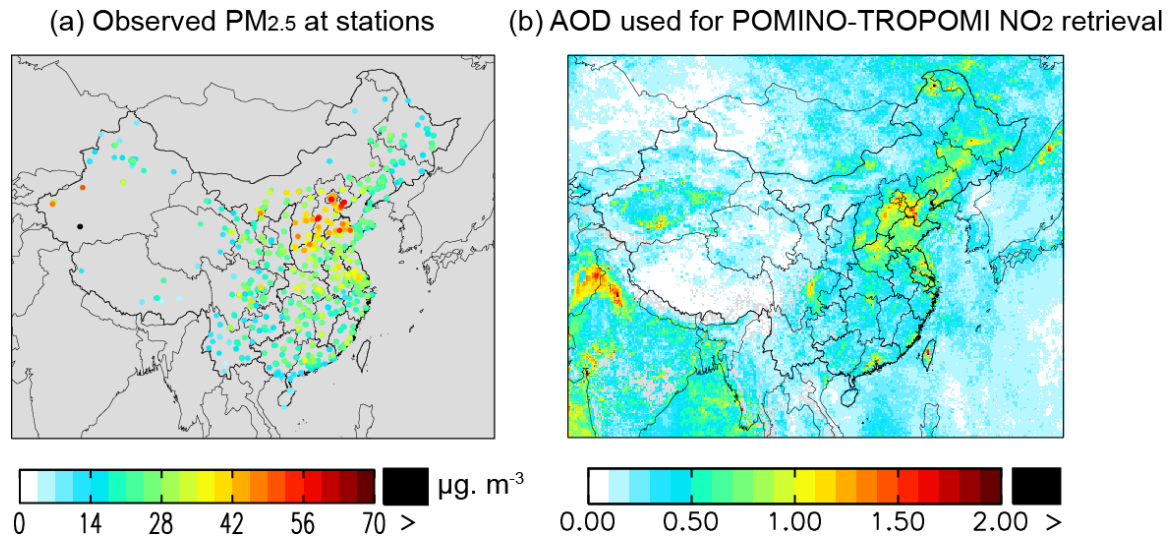
133 where I_{cld} denotes the radiance from the cloudy part of the pixel, I_{clr} the radiance
134 from the clear-sky part of the pixel, f_{eff} the cloud fraction (CF), and the R the total
135 scene radiance. Retrieval of cloud properties is a prerequisite for NO_2 retrieval. We
136 take the effective cloud pressure (CP) from the FRESKO-S algorithm (van Geffen et
137 al., 2019) which uses the O_2 A-band (around 758 nm) for TROPOMI trace gas
138 retrievals. We re-calculate w and f_{eff} at the NO_2 fitting wavelength (437.5 nm). The
139 online CF calculation is similar to that for TM5-MP-DOMINO (Arnoud et al., (2017);
140 van Geffen et al., 2019), but with an explicit aerosol correction to be consistent with
141 the following NO_2 retrieval and to remove the aerosol signal from the retrieved CF
142 data.

143 For explicit aerosol corrections in this study, we take daily aerosol simulation results
144 from the GEOS-Chem v9-02 nested model over East Asia, followed by a monthly
145 AOD correction using MODIS/Aqua C6.1 AOD data. Our future study will use the
146 CALIOP aerosol extinction vertical profiles to further improve the modeled aerosol
147 profiles. Figure 2b shows the spatial distribution of AOD in July 2018 used in clouds
148 and NO_2 retrievals. The AOD distribution is consistent ($R = 0.42$, $N = 1447$) with that
149 of near-surface $\text{PM}_{2.5}$ mass concentration measurements (Fig. 2a) taken from the
150 Ministry of Ecology and Environment of China (MEE); the difference between AOD
151 and near surface $\text{PM}_{2.5}$ is expected because they represent different parameters of
152 aerosols.

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

158 In addition to our formal POMINO-TROPOMI product (referred to as Case REF), we
159 use sensitivity cases to evaluate the impacts of aerosol corrections (explicit versus
160 implicit) and the horizontal resolution of a priori NO_2 profiles (Cases 1 and 2 in Table
161 2). Two additional cases (Cases 3 and 4) concern the treatment of CP in combination

162 with the choice of aerosols and surface reflectance. Specifically, using the CP data
163 directly from FRESCO-S means that our retrieval algorithm does not perfectly
164 account for the effect of aerosols on clouds. Our retrieval consider the BRDF effects
165 while Lambertian surface is used in deriving the FRESCO-S CP. To ensure sampling
166 consistency, the pixels used in all cases are selected based on the CRF values in Case
167 REF.



168 Figure 2. (a) Observed near-surface PM_{2.5} mass concentrations averaged over July
169 2018. Results are sampled at the times of valid TROPOMI data. (b) AOD data on a
170 0.05° × 0.05° grid used for the retrieval of POMINO-TROPOMI NO₂ VCDs in July
171 2018.

172 2.2 Ground-based MAX-DOAS measurements

173 We use ground-based MAX-DOAS measurements to validate the
174 POMINO-TROPOMI NO₂ product. The MAX-DOAS measurements are from two
175 suburban stations (Xuzhou and Nanjing) and one remote station (Fukue, (Kanaya et
176 al., 2014)). Table 1 shows the geographical and time information of each
177 MAX-DOAS site, and SI Part A describes each instrument in detail. Although
178 Xuzhou and Nanjing are both classified as suburban sites located at university
179 campuses, the NO₂ spatial distributions around the two sites are very different. The
180 spatial distribution of NO₂ VCDs is relatively smooth around Xuzhou, whereas the
181 VCDs exhibit a strong spatial gradient around Nanjing (Figure 3).

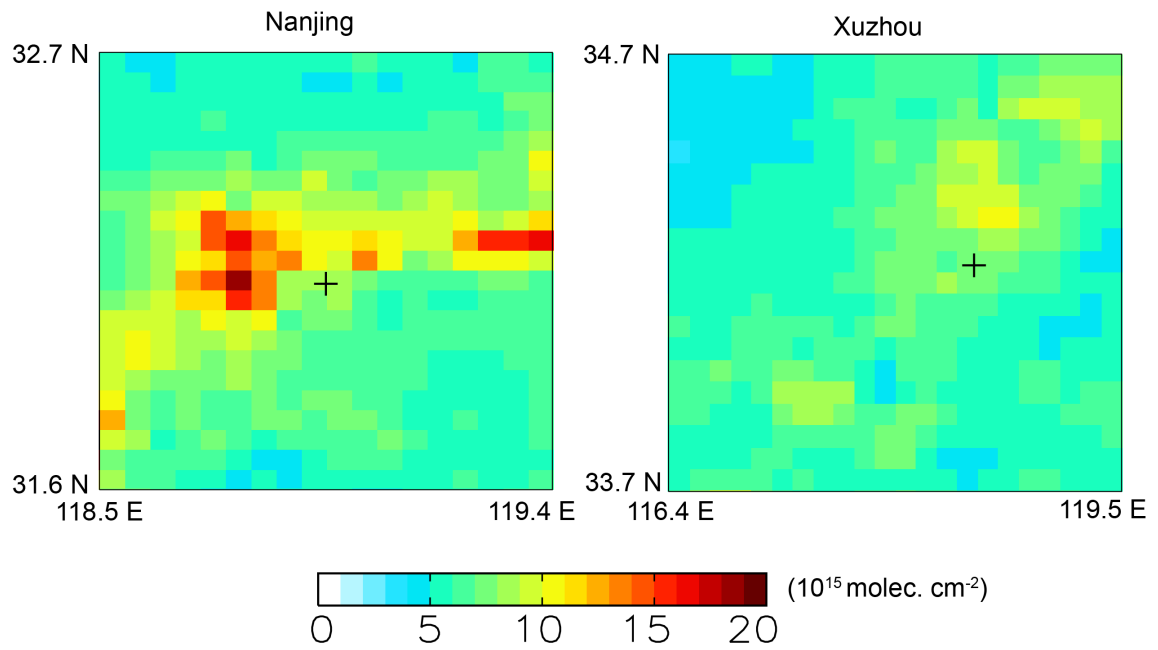
182 A consistent spatiotemporal sampling is crucial in comparing satellite measurements
183 and MAX-DOAS data (Boersma et al., 2018; Lin et al., 2014; Liu et al., 2019; Wang
184 et al., 2017). We average all valid MAX-DOAS data within ± 1 hours of the
185 TROPOMI overpass time to obtain daily values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

190

191 Table 1. MAX-DOAS measurement sites.

Site name	Geographical location	Measurement time
Nanjing	118.950° E, 32.118° N, 36 m	2018/07/01-2018/10/31
Xuzhou	117.142° E, 34.217° N, 92 m	2018/07/01-2018/10/31
Fukue	128.680° E, 32.750° N, 83 m	2018/07/01-2018/09/15

192



193 Figure 3. Spatial distributions of POMINO-TROPOMI NO₂ VCDs (on a 0.05° ×
 194 0.05° grid) around (a) Nanjing and (b) Xuzhou MAX-DOAS measurement sites in
 195 July 2018. The MAX-DOAS sites are marked as “+”.

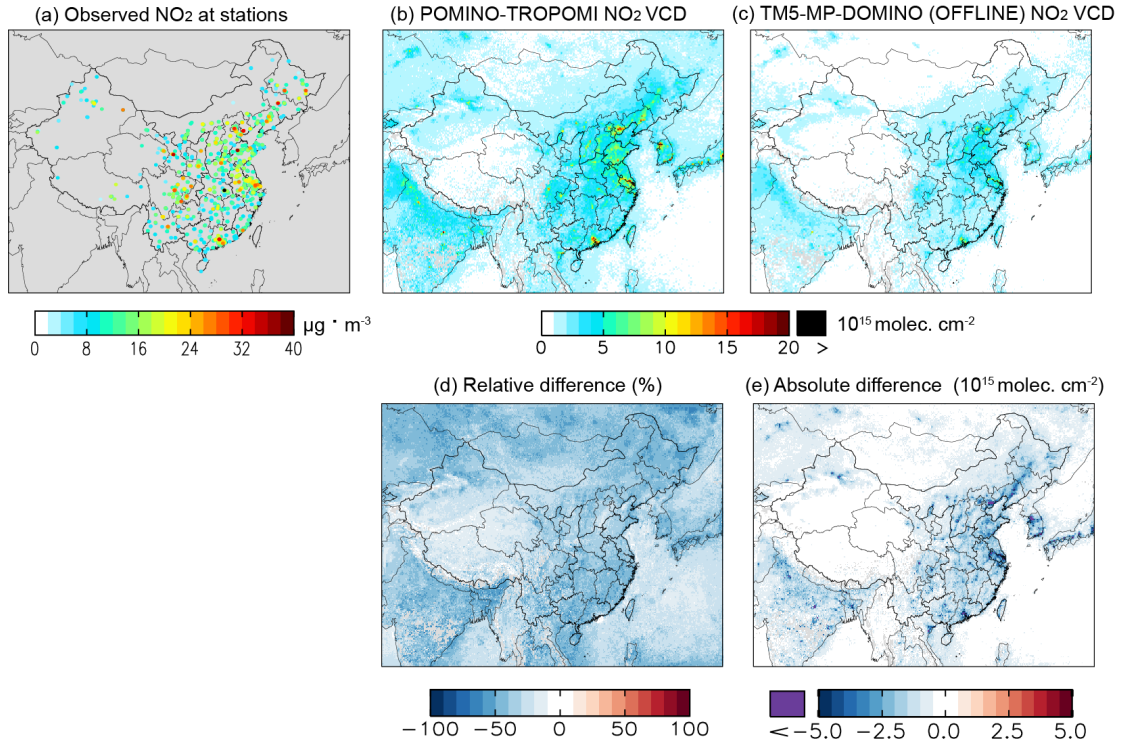
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197 **3 Results**

198 *3.1 POMINO-TROPOMI NO₂ VCDs over East Asia*

199 Figure 4b shows the spatial distribution of POMINO-TROPOMI tropospheric NO₂
200 VCDs over East Asia on a 0.05° × 0.05° grid in July 2018. High VCD values (> 3
201 × 10¹⁵ molec. cm⁻²) are shown over polluted areas such as East China and India, and
202 low values (< 1 × 10¹⁵ molec. cm⁻²) over the open ocean and the Tibetan Plateau.
203 For comparison, the colored dots in Fig. 4a visualize the near-surface NO₂
204 concentrations observed at the MEE sites at the overpass time of TROPOMI. In both
205 the VCD and the near-surface concentration maps (Fig. 4a and b), hotspots like urban
206 centers and isolated sources can be seen clearly, due to the short lifetimes of NO_x in
207 summer. The spatial correlation is about 0.55 (N = 1458) between the VCD and the
208 near-surface concentration distributions.

209 Figure 4c shows the spatial distribution of TM5-MP-DOMINO (OFFLINE) NO₂
210 VCDs for comparison. The general distribution of TM5-MP-DOMINO is consistent
211 with that of POMINO-TROPOMI with a correlation coefficient of 0.97 (N =
212 1091154). However, TM5-MP-DOMINO NO₂ VCD values are lower than
213 POMINO-TROPOMI by about 35% averaged over the whole domain (Fig. 4d), by
214 -37 - 68% over cleaner areas (POMINO-TROPOMI < 5 × 10¹⁵ molec. cm⁻²), and
215 by 0 - 66% over more polluted areas (POMINO-TROPOMI ≥ 5 × 10¹⁵ molec.
216 cm⁻²). TM5-MP-DOMINO does not show strong local signals at pollution hotspots
217 like the urban center of Beijing (Fig. 4c). Over these hotspot locations,
218 TM5-MP-DOMINO is lower than POMINO-TROPOMI by up to 5 × 10¹⁵ molec.
219 cm⁻².



220

221 Figure 4. The spatial distribution of (a) NO_2 at monitoring stations, (b) AMFv6
 222 derived NO_2 VCD and (c) TM5-MP-DOMINO (OFFLINE) NO_2 VCD at $0.05^\circ \times$
 223 0.05° grid in July 2018. (d) and (e) are relative and absolute difference of
 224 TM5-MP-DOMINO (OFFLINE) to POMINO-TROPOMI NO_2 VCD.

225

226

227 Table 2. Sensitivity experiments for the NO_2 retrieval based on the
 228 POMINO-TROPOMI algorithm.

ID	A priori NO_2 profiles	Aerosols	Surface reflectance
Case REF (POMINO-TROPOMI)	0.25° lat. \times 0.3125° long.	Explicit	MODIS BRDF
Case 1	Same as Case REF	N/A	Same as Case REF
Case 2	2° lat. \times 2.5° 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

229

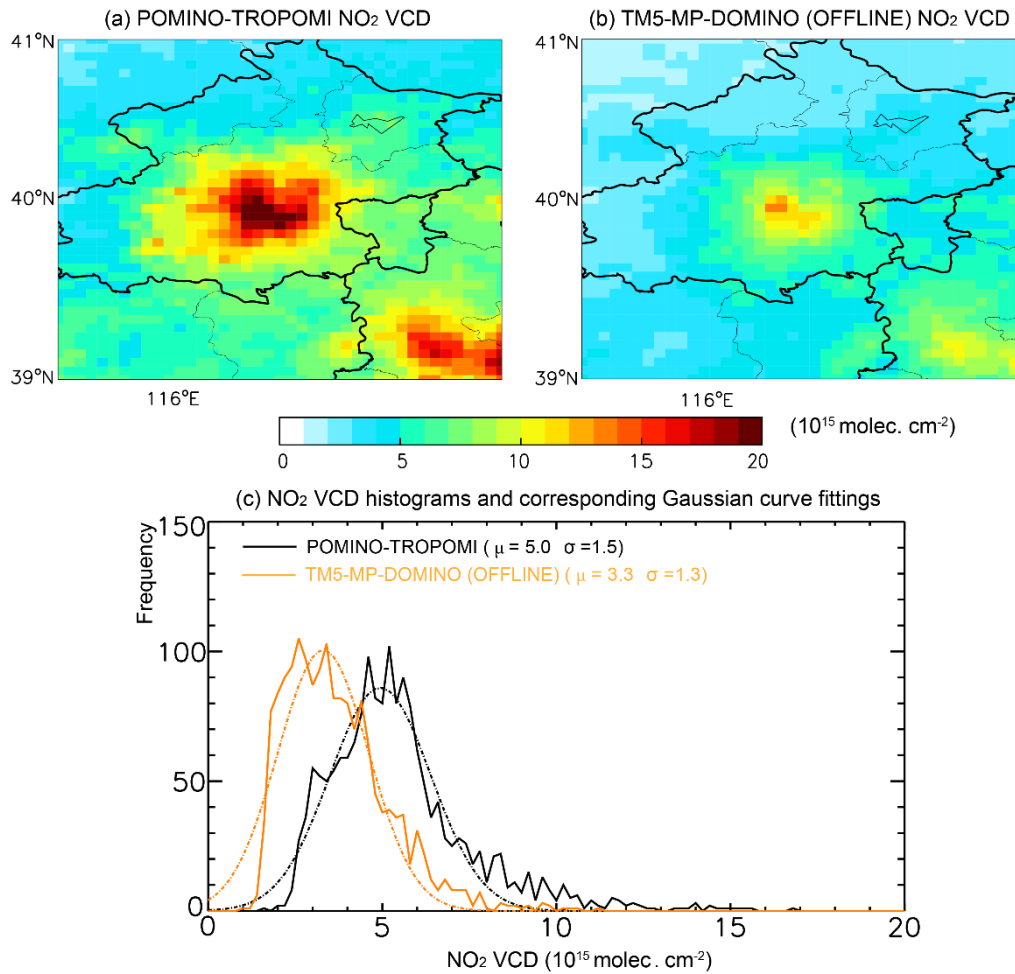
¹ Explicit aerosol treatments for M_{clr} and no aerosol corrections for M_{cld} .

230

² The LER dataset is a five-year climatology built upon OMI measurements on a grid of $0.5^\circ \times 0.5^\circ$. The dataset is

231

taken from the TM5-MP-DOMINO product.



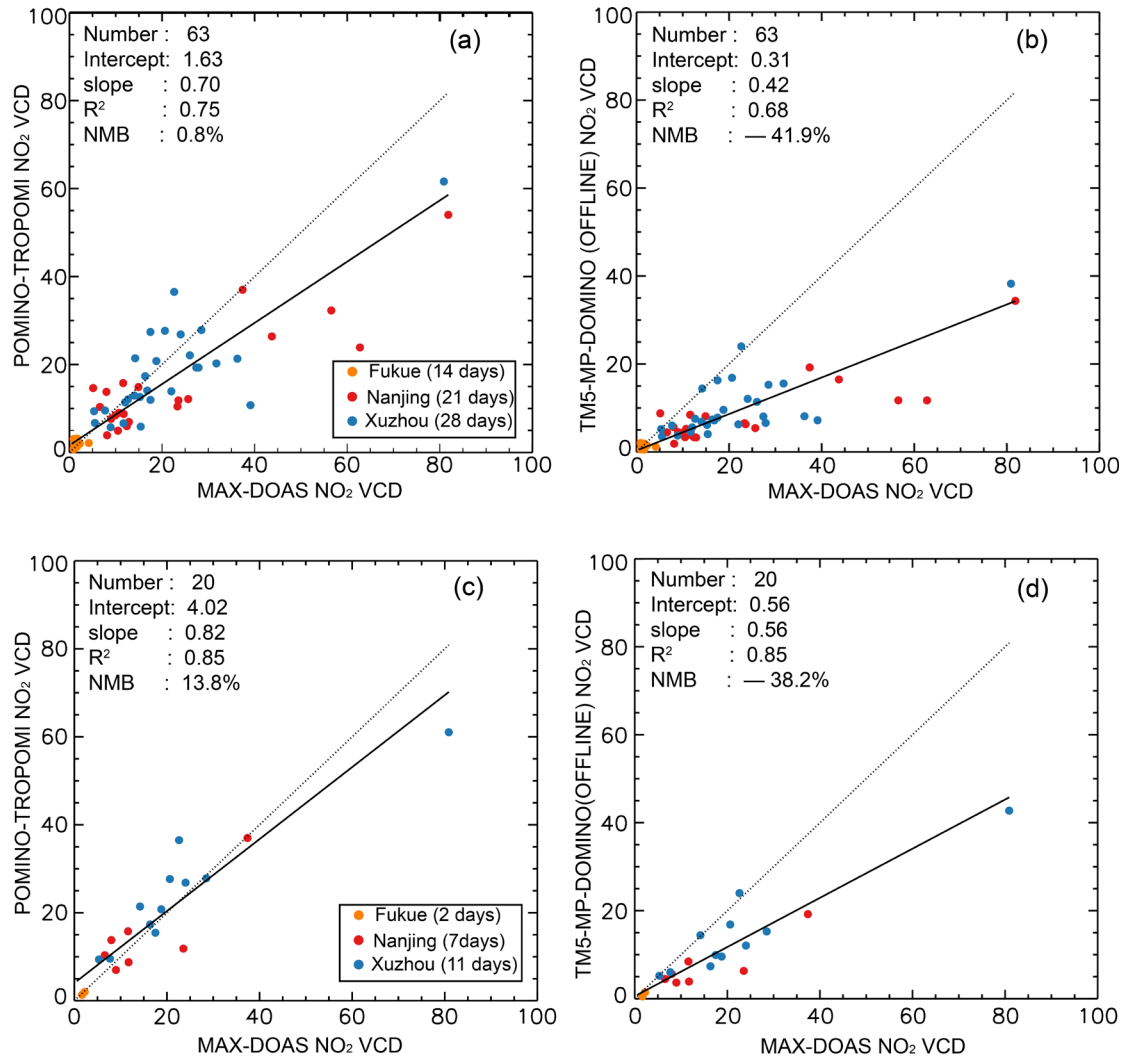
232 Figure 5. Spatial distributions of (a) POMINO-TROPOMI NO₂ VCDs and (b)
 233 TM5-MP-DOMINO (OFFLINE) NO₂ VCDs on a 0.05° × 0.05° grid over Beijing
 234 and its surrounding areas averaged over July 2018. (c) Histograms of monthly NO₂
 235 VCDs over this region. The bin size is 0.2 × 10¹⁵ molec·cm⁻². The black and orange
 236 dashed lines are corresponding Gaussian curve fitting of the histograms. μ is the
 237 mean value and σ is the standard deviation of a Gaussian curve fitting.

238 Figure 5a and b present the two products over Beijing and surrounding areas, showing
239 a much weaker spatial gradient of NO₂ VCDs from Beijing urban center to its
240 outskirts in TM5-MP-DOMINO than in POMINO-TROPOMI. The corresponding
241 histograms and Gaussian fittings in Fig. 5c also show a lower mean value and a
242 smaller standard deviation of TM5-MP-DOMINO than POMINO-TROPOMI. These
243 results highlight the important differences between the two products at fine scales.

244 We further compare the two satellite products with ground-based MAX-DOAS
245 measurements at three sites. The scatter plots in Fig. 6a and b compare the NO₂ VCDs
246 on 63 days (from 109 pixels) over July–October 2018 with their MAX-DOAS
247 counterparts. Different colors differentiate the sites. POMINO-TROPOMI captures
248 the day-to-day variability in MAX-DOAS data ($R^2 = 0.75$) and shows a small
249 normalized mean bias (NMB = 0.8%). The reduced major axis (RMA) regression
250 shows a slope of 0.70, mainly because of the underestimate on high-NO₂ days.
251 TM5-MP-DOMINO is also correlated with MAX-DOAS ($R^2 = 0.68$), although the
252 correlation is weaker than our retrieval. The NMB of TM5-MP-DOMINO is much
253 more significant (−41.9%) and the RMA regression slope is much smaller (0.42).
254 Similar underestimates of TM5-MP-DOMINO have been discussed in their Readme
255 document (Eskes et al., 2019) and ATBD file (van Geffen et al., 2019) in general, in
256 Griffin et al. (2019) for Canada. Major plausible causes of the underestimate of
257 TM5-MP-DOMINO include coarse-resolution climatological surface albedo data,
258 coarse-resolution ($1^\circ \times 1^\circ$) a priori NO₂ profiles, implicit aerosol corrections, and
259 uncertainties of CP from FRESCO-S.

260 Cloud pressures from FRESCO-S are found to be too high, i.e., the cloud top is too
261 close to the ground, especially over China (van Geffen et al., 2019). We examine this
262 effect by excluding the pixels with CP > 850 hPa when comparing with MAX-DOAS
263 data. With this additional criterion, the number of valid days drop to 20. Figure 6c and
264 d show the scatter plots and corresponding RMA results. The NMB of
265 TM5-MP-DOMINO is reduced slightly to -38.2%, and its R^2 for day-to-day variation
266 is increased from 0.68 to 0.85. POMINO-TROPOMI still outperforms
267 TM5-MP-DOMINO: 0.85 versus 0.85 for R^2 , 13.8% versus -38.2% for NMB, and
268 0.82 versus 0.56 for RMA regression slope. The improvement from excluding CP >
269 850 hPa scenes is larger in TM5-MP-DOMINO (with implicit aerosol corrections)
270 than in POMINO-TROPOMI (with explicit aerosol corrections). The averaged CF of
271 data excluding CP > 850 hPa is 0.13 (AOD = 0.63), which is much larger than the
272 averaged value (CF = 0.06) in Fig. 6a and b (AOD = 0.57). This appears to imply that
273 the overestimated CP be partly because the FRESCO-S cloud algorithm might
274 mis-interpret heavy aerosol loadings near the ground as clouds, a common issue in
275 satellite data (Lin & Li, 2016).

276



278 Figure 6. Scatter plot for daily NO₂ VCDs (10^{15} molec·cm⁻²) between MAX-DOAS
 279 and two TROPOMI NO₂ VCD products. Each colored dot represents a day and each
 280 color denotes a station. For each day, the satellite data are averaged over all pixels.
 281 (c) and (d) are results for the two TROPOMI NO₂ VCD products with effective cloud
 282 pressures \leq 850 hpa.

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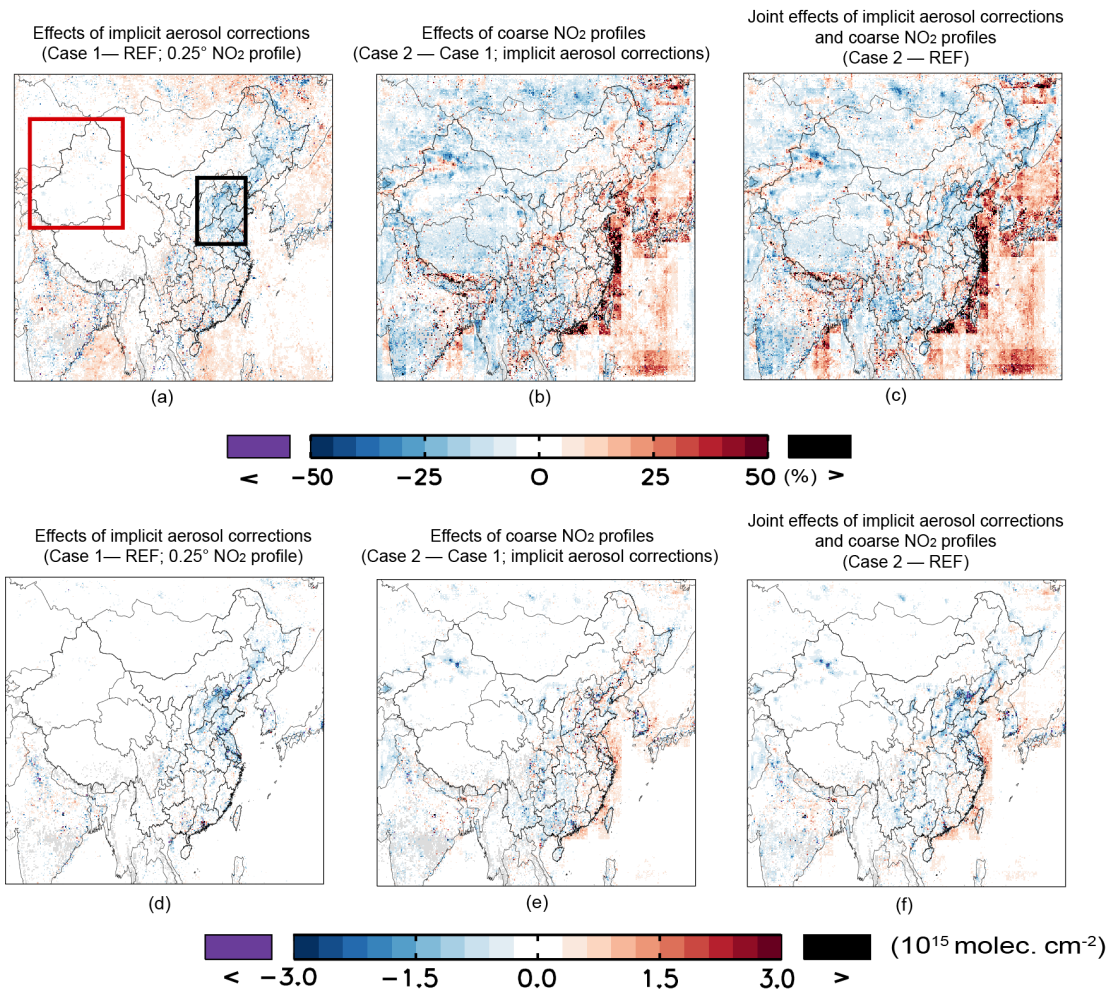
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286 3.2 Influences of aerosol correction approaches and horizontal resolutions of a priori
287 NO_2 profiles

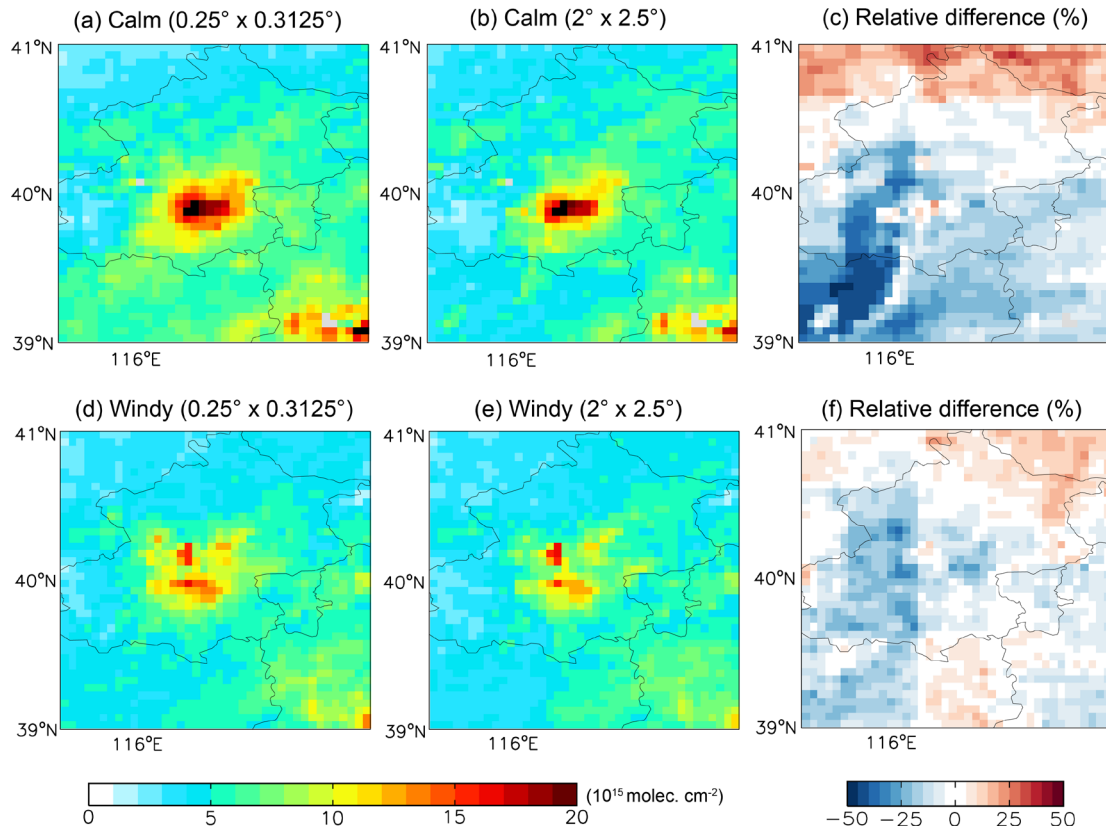
288 To investigate the causes of difference between POMINO-TROPOMI and
289 TM5-MP-DOMINO (OFFLINE), we conduct two sensitivity retrievals based on the
290 POMINO-TROPOMI algorithm (Cases 1 and 2 in Table 2). Figure 7 shows the
291 relative (a–c) and absolute (d–f) differences between retrieval cases (REF, Case 1
292 and Case 2). Case 1 adopts the implicit aerosol correction for both clouds and NO_2
293 retrievals, as in TM5-MP-DOMINO, so the difference between REF and Case 1
294 indicates the effect of aerosol representation (explicit versus implicit) (Fig. 7a, d). In
295 Case 2, the high-resolution (0.25° lat. \times 0.3125° long.) NO_2 profiles are replaced
296 by low-resolution (2° lat. \times 2.5° long.) profiles simulated by the GEOS-Chem
297 global model; aerosols are represented implicitly as in Case 1. Case 2 thus mimics
298 TM5-MP-DOMINO, which uses an implicit aerosol correction and coarse-resolution
299 NO_2 profiles. Thus, the difference between Case 1 and Case 2 arises from the a priori
300 NO_2 profiles (Fig. 7b, e). The difference between Case 2 and REF further indicates
301 the joint effect of using coarse-resolution a priori NO_2 profiles and an implicit aerosol
302 correction (Fig. 7c, f).

303 Figure 7 shows that the individual influences of aerosol representations (explicit
304 versus implicit) and a priori NO_2 profiles (fine versus coarse resolution) vary
305 substantially from one location to another. The impacts of aerosol corrections are
306 most evident over the areas of heavy aerosol loadings including East China, India and
307 parts of Southeast Asia. The implicit aerosol correction (Case 1) tends to result in
308 lower NO_2 VCDs by 0-50% over urban areas compared to the explicit aerosol
309 correction (Case REF) (Fig. 7a, d). By comparison, the impacts of NO_2 profiles are
310 spatially more heterogeneous (Fig. 7b and e). Over the offshore coastal areas, using
311 coarse-resolution NO_2 profiles (Case 2) tends to overestimate the NO_2 VCDs by
312 30-100% relative to when high-resolution profiles are used (Case 1), due to overly
313 (horizontal) smoothing of NO_2 .

314 Below, we discuss these differences over two key areas including Northern East
315 China and Xinjiang. Northern Eastern China (bounded by the black rectangle in Fig.
316 4a) is a heavy aerosol-loaded region (Fig. 2). Over this region, an implicit rather than
317 explicit aerosol representation results in lower NO_2 VCDs by $\sim 25\%$ (Fig. 7a, d), while
318 the effect of NO_2 profiles is weaker (Fig. 7b, e). The joint effect is dominated by the
319 effect of aerosol representation (Fig. 7c, f).



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



323 Figure 8. Spatial distributions of POMINO-TROPOMI NO₂ VCDs on a 0.05° ×
 324 0.05° grid under calm conditions with (a) 0.25° lat. × 0.3125° long. NO₂ profiles
 325 (Case 1), with (b) 2° lat. × 2.5° long. profiles (Case 2), and (c) their relative
 326 differences. (d)–(f) is similar to (a)–(c) but under southward wind.
 327

328 That the impact of a priori NO₂ profiles is relatively small over Northern Eastern
329 China is partly because of the smearing effect by wind. Figure 8 differentiates the
330 effects of NO₂ profiles between twenty-three windy (daily average wind speed under
331 500 m > 2 m s⁻¹) days and seven calm (wind speed < 2 m s⁻¹) days. The dataset of
332 wind is taken from National Aeronautics and Space Administration/Global Modeling
333 and Assimilation Office's (NASA/GMAO's) "forward-processing" (GEOS-FP) data
334 product with the horizontal resolution at 0.25° lat. × 0.3125° long. In the windy
335 cases, the NO₂ VCDs are much more smoothed even at 0.3125° resolution, thus the
336 difference of NO₂ resolutions is very small. In contrast, NO₂ VCDs exhibit much
337 stronger horizontal gradient in calm situations, which are better retrieved when
338 high-resolution NO₂ profile are used. As calm situation is more favorable for pollutant
339 accumulation while windy days help to dilute concentration of NO₂, so much higher
340 NO₂ VCDs are found in Fig. 8a and b.

341 Xinjiang (bounded by the red rectangle in Fig. 7a) is a deserted region in West China.
342 Over Xinjiang, the resolution of a priori NO₂ profiles affects the retrieved NO₂ VCDs
343 much more than the aerosol representation does (Fig. 7b and e). Compared to Case 1
344 (with high-resolution NO₂ profiles), Case 2 (with coarse-resolution profiles) leads to
345 much lower NO₂ VCDs over the isolated urban areas which are not resolved by the
346 coarse model.

347 *3.3 Influences of directly using the CP data from FRESCO-S*

348 As we take the CP data directly from the FRESCO-S retrieval rather than re-retrieving
349 CP (as done for CF), two main issues arise. First, the FRESCO-S retrieved CP may be
350 affected by aerosols, thus using such CP data in our explicit aerosol corrections (Case
351 REF) may lead to over-correction of aerosol effects. To estimate the effect of such
352 over-correction on retrieved NO₂ VCDs, we employ in an additional sensitivity case
353 (Case 3 in Table 2) a "semi-explicit" aerosol correction approach. This approach
354 explicitly includes aerosols in the calculation of AMF for the clear-sky portion (M_{clr})
355 of a pixel (as in Case REF), but excludes aerosols for the cloudy-sky portion (M_{cld}) of
356 that pixel. Correspondingly, CF is re-calculated on the basis that the radiance at 437.5
357 nm received by TROPOMI is contributed from the aerosol-contained clear-sky
358 portion and the no-aerosol cloudy-sky portion. Table 3 shows that in July 2018, on a
359 pixel basis, the derived NO₂ in Case 3 are larger than those in Case REF, with an
360 average difference increasing from 3.1% at relatively clean situations (NO₂ VCDs in
361 Case REF < 5×10^{15} molecu·cm⁻²) to 11.2% for polluted situations (NO₂ VCDs in
362 Case REF ≥ 15×10^{15} molecu·cm⁻²). The spatial distributions in Fig. S1a and S1b
363 also show higher NO₂ VCDs in Case 3 relative to Case REF. The corresponding
364 increases in CF (Fig. S1c versus S1d) are because in Case 3 the scattering
365 contributions to the radiance from aerosols in the cloudy-sky portion (that would have

366 occurred) are accounted for with higher CFs. The enhanced “shielding effect” of
 367 clouds (due to higher CFs) result in lower NO₂ AMFs and higher VCDs.

368 For surface reflectance, Case REF considers the BRDF effect instead of Lambertian
 369 Equivalent Reflectivity (LER) which is used in deriving FRESCO-S clouds. This
 370 leads to inconsistency between when the CP is derived and when it is used. The LER
 371 data used by FRESCO-S are generated at 758 and 772 nm (based on the Global
 372 Ozone Monitoring Experiment-2), rather than at the 437.5 nm for the NO₂ retrieval.
 373 Thus Case 4 adopts the OMI LER data from TM5-MP-DOMINO, a five-year monthly
 374 based climatology at 440 nm, and re-calculates CFs and NO₂ AMFs with explicit
 375 aerosol corrections and FRESCO-S CP (Table 2). Here, the ice-snow flag in the
 376 TM5-MP-DOMINO product is used to exclude the possible ice/snow contamination,
 377 and only the pixels with blue-sky albedos (derived from the BRDF data in Case REF)
 378 less than 0.3 are taken into consideration. The resulting NO₂ VCDs in Case 4 are
 379 lower than Case REF by 3.7% on pixel-based average for relatively clean situations
 380 and by 8.3% for polluted situations (Table 3). Figure S2a and S2b further shows the
 381 spatial distributions of the relative and absolute differences in derived monthly mean
 382 NO₂ VCDs between Case 4 and Case REF. In general, Case 4 leads to lower NO₂
 383 VCDs than Case REF because of stronger surface reflectance, as is obvious in the
 384 comparison of blue-sky albedo in Case REF and LER albedo in Case 4 (Fig. S2c
 385 versus S2d).

386

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

Situation ¹	Effect of aerosol correction choice (Case 3 – REF) ²	Effect of surface reflectance choice (Case 4 – REF) ²
NO ₂ VCD < 5 × 10 ¹⁵ molecu·cm ⁻² Mean AOD: 0.27 Mean CF: 0.07 Mean CP: 748 hPa	3.1%	–3.7%
5 × 10 ¹⁵ molecu·cm ⁻² ≤ NO ₂ VCD < 15 × 10 ¹⁵ molecu·cm ⁻² Mean AOD: 0.65 Mean CF: 0.08 Mean CP: 767 hPa	4.1%	–8.1%
15 × 10 ¹⁵ molecu·cm ⁻² ≤ NO ₂ VCD Mean AOD: 0.66 Mean CF: 0.09 Mean CP: 772 hPa	11.2%	–8.3%

389 ¹ The values of AOD, CF and CP shown here are mean values of the pixels of corresponding subsets in Case REF.

390 ² The percentage values represent the mean relative differences relative to Case REF.

391 *3.4 Error estimates for POMINO-TROPOMI NO₂ AMFs*

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

408

409

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

Error source		Estimated error in parameter	Corresponding error in tropospheric NO ₂ AMF
NO ₂ profiles			±10% ¹
Cloud fraction		±0.01 ²	±10% ²
Cloud pressure		±50 hPa ³	±10% ³
BRDF coefficients		±10% ⁴	±10% ⁴
Aerosol	AOD	±0.07 ⁵	±10% for clean situations; Up to ±20% for heavy-polluted cases ⁸
	SSA	±0.03 ⁶	
	Aerosol layer height	Underestimated by 300–600 m ⁷	

411 ¹ This error estimate is based on Lin et al. (2010), Lin et al. (2014), Boersma et al. (2004, 2011, 2019) and this
 412 study. The error accounts for effect of horizontal resolutions and the vertical process in GEOS-Chem.

413 ² This accounts for the expectation that our explicit aerosol correction leads to more reasonable CFs.

414 ³ Based on van Geffen et al. (2019). Our estimated NO₂ errors related to the use of FRESCO CP data (instead of
 415 re-calculating it) are within this error range.

416 ⁴ The estimated parameter error is taken from Zhou et al. (2010), and the corresponding AMF error is based on Lin
 417 et al. (2014) and Case 4.

418 ⁵ Based on Tian et al. (2019), who compared the MODIS Merged AOD C6.1 product with AERONET in China.

419 ⁶ Based on Lin et al. (2015), who compared GEOS-Chem simulated SSA with Lee et al. (2007).

420 ⁷ Based on Liu et al. (2019), who compared GEOS-Chem simulated aerosol extinction profiles with CALIOP.

421 ⁸ This is a tentative error estimate based on Lin et al. (2015), Lorente et al. (2017), Liu et al. (2019) and this study.
422 With explicit aerosol corrections, the errors in heavy-polluted situations are expected to be smaller than when
423 assuming implicit aerosol corrections.
424

425 **4 Conclusion and Discussion**

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

434 Over heavy aerosol-loaded regions, the accuracy of retrieved NO₂ VCDs is affected
435 substantially by how aerosols are represented in the retrieval process (implicit or
436 explicit). The implicit aerosol representation tends to underestimate the NO₂ VCDs by
437 0-50% over most urban areas in East Asia and by about 25% over Northern East
438 China. Using a priori NO₂ profile data at a horizontal resolution of ~ 25 km,
439 POMINO-TROPOMI captures the city-scale NO₂ hotspots. Reducing the horizontal
440 resolution of a priori profiles to what is typically set up by global models (100–200
441 km) underestimates the NO₂ hotspots and the spatial gradient surrounding them, and
442 the effects are more pronounced in calm than in windy situations. Overall, our results
443 provide useful information to improve TROPOMI retrieval algorithms, and offer
444 insight for applications to the upcoming geostationary satellite instruments including
445 GEMS, TEMPO and Sentinel-4 Precursor.

446 Further work can be done to improve the retrieval algorithm for TROPOMI. The
447 hybrid cloud retrieval method used in POMINO-TROPOMI is not optimal, because
448 only cloud fraction but not cloud pressure is re-calculated with explicit aerosol
449 corrections and BRDF effects. The uncertainty caused by inconsistent assumptions of
450 aerosols and albedos in cloud pressure and NO₂ VCD retrievals are initially estimated
451 in this study. Re-calculation of cloud pressure will be done in the future using the
452 O₂-O₂ method when the O₂-O₂ SCD data are available. Second, correcting the
453 simulated aerosol extinction vertical profiles will further improve the clouds and NO₂
454 retrievals (Liu et al., 2019). Third, the horizontal resolution of a priori NO₂ profiles (~
455 25 km) does not match the fine footprint of TROPOMI, and further increasing the
456 resolution will help achieve better accuracy for analyses of very fine scale pollution
457 characteristics such as along the highways and rivers and within urban centers.

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462 *Data availability.* The POMINO-TROPOMI NO₂ data are available at our website
463 (<http://www.pku-atmos-acm.org/acmProduct.php/>). The TM5-MP-DOMINO product
464 can be download via TEMIS website: <http://www.temis.nl/airpollution/no2.html>. The
465 near surface data of NO₂ and PM_{2.5} can be downloaded from:
466 <http://www.cnemc.cn/sss/cskqzl/>. The ground-based MAX-DOAS observations
467 would be provided after the applications of users are approved by corresponding
468 owners.

469
470 *Author contributions.* J.-T. Lin conceived the research. M.-Y. Liu and J.-T. Lin
471 designed the experiment. M.-Y. Liu performed all calculations with some code support
472 from H. Kong and H.-J. Weng. M.-Y. Liu and J.-T. Lin wrote the paper with inputs
473 from K.-F. Boersma and J.-H. Eske. R. Spurr provided LIDORT. Y. Kanaya, Q. He, X.
474 Tian, K. Qin, P.-H. Xie provided the MAX-DOAS observations. R.-J. Ni helped to
475 process surface observations. Y.-Y. Yan and J.-X. Wang helped to analyze the
476 relationship between meteorological and NO₂ VCD variations. All authors commented
477 on the manuscript.

478
479 *Competing interests.* The authors declare no conflicts of interest.
480

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