

Retrieval of tropospheric NO2 columns from satellite measurements in presence of cirrus: A theoretical sensitivity study using SCIATRAN and prospect application for the A-Train

Jérôme Vidot, Olivier Jourdan, Alexander Kokhanosvky, Frédéric Szczap, Vincent Giraud, Vladimir Rozanov

To cite this version:

Jérôme Vidot, Olivier Jourdan, Alexander Kokhanosvky, Frédéric Szczap, Vincent Giraud, et al.. Retrieval of tropospheric NO2 columns from satellite measurements in presence of cirrus: A theoretical sensitivity study using SCIATRAN and prospect application for the A-Train. Journal of Quantitative Spectroscopy and Radiative Transfer, 2010, 111 (4), pp.586-601. $10.1016/j.jqsrt.2009.10.015$. hal-01971909ff

HAL Id: hal-01971909 <https://hal.science/hal-01971909v1>

Submitted on 20 Mar 2022

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

Abstract 21

22A theoretical sensitivity study of the influence of cirrus cloud properties on tropospheric $NO₂$ 23 columns retrieved from the spaceborne Ozone Monitoring Instrument (OMI) measurements is 24 performed. It is conducted within the framework of the synergetic use of A-Train sensors to

derive more representative trace gas products. We aim to study the potential effects of cirrus 1 2clouds on tropospheric $NO₂$ retrievals using a retrieval algorithm that, unlike the OMI Standard and DOMINO algorithms, does not correct for the effects of clouds. The sensitivity 3 4study is based on the radiative transfer code SCIATRAN that performs both simulations of 5top of atmosphere (TOA) reflectances as measured by an OMI-like band and tropospheric $6NO₂$ column retrievals based on the Differential Optical Absorption Spectroscopy (DOAS) method. The results of the sensitivity study show that if a correction for cirrus clouds is not 7 8included in our simple retrieval that does not account for clouds in the first place, the 9tropospheric column can be underestimated by 55%. This underestimation depends strongly 10on cirrus parameters as, in order of importance, cloud fraction, cloud optical depth, 11asymmetry factor of cirrus cloud phase function and cloud top height. The perspective of the 12synergy between OMI and cloud information obtained from cloud-derived products of the A-13Train is evaluated in two parts by applying a simple cloud correction scheme based on the 14independent pixel approximation (IPA). Firstly, we evaluated the tropospheric $NO₂$ column 15 retrievals error caused by uncertainties in cirrus cloud properties. Secondly we studied the 16influence of subpixel cloud optical depth variability on $NO₂$ retrievals. From our simulations, 17it is demonstrated that the error will be reduced significantly if the cloud fraction is lower or 18 equal to 0.5. In this case, the cloud fraction and the cloud optical depth must be known within 19 accuracy less than 0.05 and 50 %, respectively. The cloud top height and the asymmetry 20factor must be known within uncertainty of at least 1 km and less than 0.05, respectively. The 21 latter result shows that the uncertainty of the asymmetry factor is a major source of error in 22the cloud correction for tropospheric $NO₂$ retrieval in the presence of cirrus.

23

24Keywords: Cirrus properties; OMI; Tropospheric NO₂ vertical column; SCIATRAN; A-25Train.

1. Introduction 11.

2

Nitrogen dioxide ($NO₂$) plays an important role in the tropospheric chemistry [1]. $NO₂$ is 4known to be one of the key species in the formation of photochemical smog during pollution 5episodes. In the troposphere, the concentration of $NO₂$ takes part in the chemical budget of 6ozone. It also contributes to acid rain and locally, to radiative forcing over industrial regions 7and urban areas [2]. To assess accurately our current knowledge of tropospheric chemistry 8 and its interaction with climate, global information about the amounts and distribution of $NO₂$ 9is required. By their high spatial and temporal coverage, satellite measurements of $NO₂$ are 10essential for air quality monitoring (for health regulation) and regional scale modelling 11(improvement of emission estimates). 3

The measurement of tropospheric $NO₂$ from space began with the precursor satellite 13Global Ozone Monitoring Experiment (GOME) [3] followed by the Scanning Imaging 14Absorption Spectrometer for Atmospheric Chartography (SCIAMACHY) [4] and the Ozone 15Monitoring Instrument (OMI) [5]. The retrieval algorithms of the $NO₂$ vertical columns are 16based on the Differential Optical Absorption Spectroscopy (DOAS) approach [6]. This 17technique is based on the analysis of differential structure of backscattered signal using non 18linear least squares fitting in a specific wavelength window. The spectral fit determines a slant 19 column density of $NO₂$ which is converted into a vertical column by application of an air mass 20factor (AMF). Retrievals of tropospheric $NO₂$ columns from GOME, SCIAMACHY and OMI 21 have demonstrated the weekly cycle of $NO₂$ [7], its relationship with NO_x emission, the 22annual trend over industrial countries [8], the global cartography of tropospheric $NO₂$ [9] and 23the diurnal evolution of $NO₂$ driven by emissions and photochemistry [10]. Space-based 24 measurements of $NO₂$ have been validated against ground-based measurements [11,12], 25atmospheric models [13,14] and aircraft measurements [9,15]. 12

Nevertheless, the retrieval of $NO₂$ from space measurements can be subject to significant 2errors. Boersma et al. [16] showed that $NO₂$ tropospheric vertical column can only be 3retrieved with an accuracy of 35-60%. The retrieval errors are dominated by uncertainties in 4the tropospheric AMF. The *a priori* NO₂ profile shape, the surface albedo and the cloud 5properties (especially cloud fraction) are the leading sources of errors associated with AMF 6computation. Moreover, clouds cover approximately 60% of the earth's surface. Hence, 7 considering the weak spatial resolutions of trace gas monitoring sensors (for GOME: 320×40 8km², SCIAMACHY: 60×30 km² and OMI: 13×24 km² at nadir), more than 90% of their 9measurements are contaminated by clouds [17]. 1

 The effects of clouds on trace gas retrieval from space measurements can be separated 11into three parts [18-22]. The first effect, called shielding effect, reduces the interaction 12between photons and trace gas underneath the cloud leading to an "apparent" decrease of 13depths of absorption lines. The second effect, called albedo effect, increases the depth of 14absorption lines of the gas layer above the cloud as compared to the clear sky. This is due to 15enhanced single and multiple scattering light paths from Sun to cloud to satellite. The third 16 effect, called in-cloud absorption, increases the depth of absorption lines of gas layer inside 17the cloud. Here, light scattering process due to the cloud is responsible for the light path 18 enhancement as compared to a nonscattering layer. These three effects are competing together 19 and the final effect depends on the cloud properties as well as on the profile of the trace gas. 20Therefore, accurate space-based retrievals of trace gas column in presence of clouds 21 necessarily imply a precise assessment of the cloud radiative and macrophysical properties 22which are, in turn, determined by their microphysical, optical and geometrical characteristics. 10

The most important point to perform a cloud correction for the retrievals of tropospheric $24NO₂$ column is to identify the cloud parameters that significantly modify the depth of 25absorption lines of $NO₂$. Studies of Boersma et al. [16] and Wang et al. [23] showed that the 23

1cloud fraction, cloud albedo and cloud pressure were important quantities for cloud 2 correction. Errors in these cloud properties will directly end up in errors in $NO₂$ columns. 3However the quantitative estimate of the retrieval errors depends on the chosen cloud model. 4For OMI, the cloud correction scheme of the current algorithm is based on the simplified Lambertian cloud model that assumes a homogeneous cloud with an albedo of 0.8 [24]. This 5 6method corrects the cloud contribution of the total reflectance at the top of atmosphere (TOA) measured for an OMI pixel using only two cloud parameters: the effective cloud fraction and 7 8cloud pressure. The use of this Lambertian cloud model compared to a scattering cloud model has been estimated to lead to AMF differences between -10% and 10% [23] and mean AMF 9 10differences of -12% with a standard deviation of 10% [25], depending on cloud properties, 11 cloud fraction and $NO₂$ pollution. However, the assumption of a Lambertian cloud model is 12only valid if the bi-directional properties of light reflectance from clouds can be neglected.

In order to overcome these limitations, the cloud correction algorithm can be improved on 14the basis of collocated data from cloud sensors and gas sensors onboard different satellites. 15The A-train satellite constellation allows for near simultaneous measurements of cloud and 16trace gas properties at different spatial scales. For instance, high spatial resolution (typically $17(1\times1 \text{ km}^2)$ cloud properties can be assessed from the Moderate Resolution Imaging 18Spectroradiometer (MODIS) [26] and from the 5-km resolution cloud product provided by the 19Cloud-Aerosol Lidar with Orthogonal Polarization (CALIOP) (Level 2 Algorithm Theoretical 20Basis Document, available at http://www-calipso.larc.nasa.gov/resources/pdfs/PC-SCI-21202_Part4_v1.0.pdf). This subpixel cloud information (cloud cover, optical thickness, cloud 22top height...) can then be used to constrain a cloud correction algorithm applied to OMI 23 measurements. Nevertheless, the use of CALIOP information is limited by the fact that 24 CALIOP only overlaps with the nadir pixel of OMI. 13

This paper lies within the framework of using the synergy of A-Train instruments to 2improve trace gas retrievals in presence of cirrus clouds. Due to their low optical thickness, 3cirrus clouds are more difficult to detect compared to cumulus, particularly over land and ice 4surfaces. Additionally, infrared measurements have shown that thin cirrus clouds (with optical 5depths between 0.1 and 1.0) have a global frequency of about 20 to 40% [27]. More recently, 6a cirrus occurrence up to 70% near the tropics over the 100° -180 $^{\circ}$ E longitude band has been 7derived from active measurements by CALIOP [28]. From these observations, this paper aims 8at studying into the potential effects of cirrus clouds on the tropospheric $NO₂$ column 9retrievals from OMI space-based measurements and the impact of uncertainties in cirrus cloud 10 properties in a cloud correction scheme based on the IPA. It is important to point out that this 11study is focusing on the effect of cirrus clouds properties in the $NO₂$ absorption band and that 12the results presented here are not compared with the operational cloud correction scheme of 13OMI that is based on the O_2 - O_2 absorption band [24]. Even if we can assume that the O_2 - O_2 14band is sensitive to cirrus, the OMI O_2-O_2 algorithm will retrieve, in presence of cirrus, a 15small value for effective cloud fraction and high cloud height. For small cloud fractions, the 16assumption on the surface albedo plays an important role and will impact the retrieval of 17tropospheric $NO₂$ column. This effect is beyond the scope of the paper, but the new surface 18albedo database made from OMI itself (Kleipool et al., 2008) will certainly change the 19tropospheric $NO₂$ retrievals, because the previous albedo database was from GOME with a 20 coarser pixel resolution than OMI. 1

Theoretical sensitivity study of the inversion of the tropospheric $NO₂$ column to 22undetected or retrieved (with errors) cirrus cloud properties is discussed in this paper. The 23 sensitivity study is based on radiative transfer model (RTM) simulations of TOA reflectances 24as measured in an OMI-like band. The first part of the paper will describe the atmospheric 25 model, the retrieval algorithm and the cirrus cloud model. In the second part of the paper, the 21

results of the sensitivity study that highlight the cirrus parameters that influence tropospheric 1 $2NO₂$ column retrievals are presented. The last part is devoted to the perspective on application 3of A-Train cloud-derived products by estimating tropospheric $NO₂$ column error caused by 4uncertainties in cirrus cloud parameter or caused by subpixel cloud optical depth 5inhomogeneity within an OMI pixel at nadir.

- 6
- 7

2. NO2 retrieval algorithm and model setups 82.

9

2.1 Atmospheric model 10

In our study, the atmosphere is treated as plane-parallel. In the spectral band used by OMI 12for NO₂ retrieval (between 405 and 465 nm), the two other trace gases, namely O₃ and H₂O 13are included with fixed standard atmospheric profiles. The collision complex of oxygen 14 molecules O_2-O_2 is included as well by assuming a fixed standard atmospheric profile of O_2 . 15The influence of cirrus clouds is evaluated in different tropospheric $NO₂$ situations 16 characterized by the three atmospheric profiles of $NO₂$ Volume Mixing Ratio (VMR in ppbv) 17shown in Fig. 1. The $NO₂$ VMR profile represented by the full line corresponds to the lowest $18NO₂$ contribution in the troposphere whereas the $NO₂$ VMR profile represented by the dotted-19 dashed line corresponds to the largest $NO₂$ contribution. The tropospheric part of the total 20 atmospheric $NO₂$ vertical column is commonly retrieved by assuming that the longitudinal 21 variation of the stratospheric $NO₂$ vertical column is small. Therefore, tropospheric $NO₂$ 22 vertical column (V_{tr}) over a polluted location (e.g., over a city) can be retrieved from the total $23NO₂$ vertical column by removing the total $NO₂$ vertical column over a clean site (e.g., Pacific 24ocean) at a given location with the same latitude. Here, we simply removed a constant 25 stratospheric $NO₂$ vertical column from the total vertical column. The stratospheric $NO₂$ 11

1vertical column was calculated as the integral of the $NO₂$ profile from 12 km to the top of 2atmosphere (TOA). Accordingly, the three profiles displayed in Fig. 1 correspond to three 3tropospheric NO₂ conditions defined as: (1) low polluted with $V_{tr} = 0.43 \times 10^{15}$ molec/cm² (full 4line), (2) moderately polluted with $V_{tr} = 2.81 \; 10^{15} \; \text{molecular/}$ (dotted line) and, (3) heavily 5 polluted with $V_{tr} = 9.36 \, 10^{15} \, \text{molecular/}$ (dotted-dashed line).

The modelling results presented in this paper are performed for single values of the 7surface albedo *A* = 0.05, the solar zenith angle $θ_s = 30°$ and the viewing zenith angle $θ_v = 0°$ 8(according to OMI nadir viewing geometry). The influence of the solar angle and the surface 9albedo are discussed in section 3. We note in passing that aerosols were chosen to be not 10included in the atmospheric model. 6

11

2.2 DOAS retrieval 12

The tropospheric $NO₂$ vertical column (V_{tr}) retrieval algorithm has been developed in the 14framework of SCIATRAN [29]. SCIATRAN is both a radiative transfer code and a retrieval 15algorithm for many atmospheric gases (see http://www.iup.uni-bremen.de/sciatran/ for full 16 description of SCIATRAN). For the retrievals of V_{tr} , the DOAS technique is applied to 17 simulated TOA reflectance in the OMI-like band with a spectral sampling of 0.21 nm [30] and 18 with a signal to noise ratio (SNR) of 1400 [10]. Retrievals of V_t are performed by converting 19the simulated TOA reflectances R_{TOA} to so-called differential optical densities D , which are 20calculated as 13

21

$$
D(\lambda) = \ln(R_{\text{TOA}}(\lambda)) - P_3(\lambda), \tag{1}
$$

23

24where $P_3(\lambda)$ is a 3rd degree least squares polynomial fit of the logarithm of R_{TOA} with respect to 25the wavelength λ , that removes slowly varying functions. The conversion of R_{TOA} into *D* 1 ensures the better contrast of the $NO₂$ absorption line depths to improve the accuracy of the 2fitting procedure. Then, in the retrieval algorithm the quadratic form

3

4
$$
F(\lambda, V_{tr,ret}) = ||D_{true}(\lambda, V_{tr, true}) - D_{ret}(\lambda, V_{tr, ret})||^2, \qquad (2)
$$

5

6is minimized with respect to unknown parameter $V_{tr,ret}$. D_{true} is the differential optical density 7spectrum simulated with SCIATRAN for the true tropospheric $NO₂$ column $V_{t, true}$. D_{ret} is the 8 retrieved differential optical density spectrum for the retrieved tropospheric NO₂ column *V*_{*tr,ret*.} 9

2.3 Cirrus cloud model 10

Cirrus clouds are treated as single layer homogeneous clouds. The most important cirrus 12 cloud parameters that potentially influence the trace gas retrieval are those which determine 13the photon paths in the atmosphere. In the visible spectral range, these parameters are 14 expected to be the cloud optical depth τ , the cloud phase function P , the cloud top height τ and 15the cloud geometrical depth *h* [31]. These parameters are used as inputs in SCIATRAN. 16Because of the quite large spatial resolution of OMI measurement (13×24 km²), one more 17important parameter is the geometric cloud fraction c . In most retrieval, c is accounted for 18 considering the independent pixel approximation (IPA). IPA consists in the hypothesis that 19 R_{TOA} is the sum of the reflectance of a cloudy part (R_{cloudy}) and the reflectance of a clear-sky 20part (R_{clear}) as 11

21

$$
R_{TOA} = cR_{cloudy} + (1 - c)R_{clear}
$$
 (3)

23

24The ice crystals phase function used for SCIATRAN simulations has been retrieved from in 25 situ aircraft measurements during the Cirrus Cloud Experiment (CIRCLE2) campaign. The

phase function (Fig. 2) was inferred from the Polar Nephelometer (PN) measurements. The 1 2best fit of the PN measurements was achieved using a combination of spherical ice particles 3with diameters ranging from 1 to 100 µm and rough droxtal shaped particules with maximum dimension between 2 and 200 µm. Accordingly, two particle size distributions were retrieved 4 5using the iterative inversion method developed by Oshchepkov et al. [32] and upgraded by 6Jourdan et al. [33]. Then, the scattering patterns of the retrieved particle size distribution were 7computed with the Lorenz-Mie theory (for spherical particles) and with an improved 8geometric-optics model (for droxtal particles) at a wavelength of 420 nm [34]. The retrieved phase function exhibits a featureless behaviour and is flat at side scattering angles which is in 9 10accordance with most of the observations [32-33,35-37] or scientific recommendations in ice 11 cloud remote sensing application $[38-41]$. The corresponding asymmetry factor g is 0.75.

To illustrate the influence of cirrus clouds on the retrieval of *Vtr*, Fig. 3 shows simulated 13differential optical densities (calculated with Eqs. 1 and 3) with and without cirrus clouds. 14 R_{clear} is calculated using SCIATRAN with standard profiles of O_3 , H_2O and O_2 , surface albedo 15 and geometries described in section 2.1. Clouds properties (τ = 1, CIRCLE2 phase function, *z* 16 = 10 km and $h = 1$ km) are added in the calculation of R_{cloudy} . The heavily polluted NO₂ profile 17 is considered. The clear-sky differential optical density ($c = 0$ in Eq. 3) is represented by the 18green line. Blue and red lines represent differential optical densities when a cirrus cloud is 19 included and cover the full pixel $(c = 1)$ and half of the pixel $(c = 0.5)$, respectively. When a 20 cirrus cloud is included, the depths of $NO₂$ absorption lines are reduced. This stems from the 21 cloud shielding effect discussed in the introduction, which is greater to both cloud albedo and 22in-cloud absorption effects. Because V_t varies with the $NO₂$ absorption lines depths, the 23 presence of undetected cirrus would lead to a lower apparent V_{tr} , that depends on the cloud 24 fraction. Nevertheless, these results are valid for algorithms that do not make any attempt to 25 correct for clouds in the first place. In the current available OMI $NO₂$ algorithms such as the 12

1OMI standard retrieval (bucsela et al, 2006) and the Dutch OMI NO₂ (DOMINO) retrieval 2(Boersma et al. 2008b), the missing tropospheric $NO₂$ column due to clouds is compensated 3by adding a ghost column obtained from chemical transport model. In situations of cirrus, the $4OMIO₂-O₂$ algorithm will retrieve a small value for effective cloud fraction and high cloud 5height. Because of the small effective cloud fraction, the ghost column effect will also be small. 6

- 7
-
- 8
- 9
-
- 10

3. Influence of undetected cirrus clouds on NO2 retrieval $113.$

12

The differential optical densities simulated previously are used to introduce the following 14 discussion that focuses on the influence of undetected cirrus cloud on the tropospheric $NO₂$ 15column retrieval. The influence of undetected cirrus clouds is evaluated as an error on the 16 retrieved tropospheric $NO₂$ column $V_{tr,ret}$ compared to a true tropospheric $NO₂$ column $V_{tr, true}$: 13 17

$$
Error = \frac{V_{tr,ret} - V_{tr,true}}{V_{tr, true}} \times 100,
$$
\n(4)

19

20V_{tr,ret} is retrieved according to the following methodology: A true TOA reflectance spectrum is 21 calculated following Eq. 3, for an atmosphere characterized by both a cirrus cloud with 22 prescribed properties and a true tropospheric $NO₂$ column $V_{tr,true}$. Then, this TOA reflectance is 23used as the input spectrum for the DOAS-based SCIATRAN retrieval algorithm. The 24 retrieved quantity is, in that case, called $V_{tr,ret}$. It is important to point out that the inversion procedure is performed with the hypothesis of a non cloudy atmosphere. By this mean, errors 1 2caused by a specific cirrus parameter (optical depth, cloud top height, etc...) can be quantified 3and used to determine the relative importance of the impact of a given cloud parameter on the 4 retrieval of tropospheric $NO₂$ columns for retrievals that do not correct for clouds in the first 5place.

6

3.1 Cloud fraction and cloud optical depth 7

The panel of Fig. 4 displays errors of the retrieved $V_{tr,ret}$ as a function of cloud optical 9depth and cloud fraction for low polluted condition (Fig. 4a), for moderately polluted 10 condition (Fig. 4b) and for heavily polluted condition (Fig. 4c). For these cases, $z = 10$ km, *h* 11 = 1 km and $g = 0.75$ (i.e., with the CIRCLE2 ice crystal phase function). The overall negative 12error (underestimation) highlights the predominance of the shielding effect. This is explained 13by the fact that the tropospheric $NO₂$ is mainly situated under the cirrus cloud. Even for the 14 highest value of the cloud optical depth considered in this study (τ = 3), the albedo effect still 15 remains negligible compared to the shielding effect. The underestimation of $V_{tr,ret}$ increases 16 rapidly with both τ and c and can reach, for $\tau = 3$ and $c = 1$, an underestimation of ~35% in 17low polluted conditions (Fig. 4a), \sim 45% in moderately polluted conditions (Fig. 4b) and ~55% in heavily polluted condition (Fig. 4c). It is interesting to notice that the 18 19underestimation remains less than 5% for optical depth less than 0.2 whatever the cloud 20fraction and the polluted conditions are. This implies, from our simulations, that optically thin 21 cirrus or subvisible cirrus do not significantly influence the $NO₂$ tropospheric column retrieval 22which is not the case for example, in atmospheric CO_2 retrieval [42,43]. This might be 23 explained by the fact that, in the $NO₂$ absorption band, the signal is predominantly due to 24 Rayleigh scattering as compared to $CO₂$ absorption band that is in the near infrared spectral 25band and where the Rayleigh scattering is very low. 8

3.2 Cloud top height and cloud geometrical depth 2

The influence of cloud top height (*z*) and cloud geometrical depth (*h*) is evaluated in term 4of error difference between the maximum and the minimum $NO₂$ tropospheric column errors 5over the different values of *z* and *h* considered. Fig. 5 displays the error caused by undetected 6 Geirrus clouds versus cloud top height *z*. In these simulations, $g = 0.75$ and $h = 1$ km. Colour lines and symbols differ with pollution conditions, cloud fraction and cloud optical depth. For 7 $8c = 1$ and $\tau = 3$ in low polluted condition (represented by the green line with triangles), the 9error is -29% at *z* = 8 km and -37% at *z* = 15 km leading to an error difference of 8% between $10z = 8$ km and $z = 15$ km. As comparison, the error difference is ~6% for moderately polluted 11 condition (represented by the blue line with triangles) and $~4\%$ for heavily polluted condition 12(represented by the red line with triangles). The underestimation increase with cloud top 13height mostly between $z = 8$ to $z = 12$ km. This is explained by the increase of shielding effect 14with the altitude of the cirrus cloud. For $c = 0.5$ and $\tau = 1$ (represented by lines with 15diamonds), error differences are almost constant with *z*. Table 1 provides a summary of error 16 differences for different values of *c* and τ . Errors remain less than 1% for low optical depth (τ 17= 0.05) or for low cloud fraction ($c = 0.05$). We noticed also that, for fixed *c* and τ , error 18 differences decrease with polluted conditions, which can be explained by an increase of the 19in-cloud absorption effect when the cloud is lower due to higher tropospheric $NO₂$ (see Fig. 1). The error caused by undetected cirrus clouds versus cloud geometrical depth *h* is 20 21 represented in Fig. 6 for the same cases as Fig. 5. Underestimation of V_t decrease slowly with *h* due to increase of in-cloud absorption leading to error difference of few percents between *h* 22 $23= 0.1$ km and $h = 2$ km. Table 2 provides a summary of error differences for different values 24 of c and τ which, overall, remain less than 3%. 3

3.3 Asymmetry factor of ice crystal phase function 1

 The influence of ice crystal phase function is evaluated by comparing the retrieval errors 3versus the asymmetry factor g . g is used to parameterize the general behaviour of the ice 4crystal phase function because it is an integrated optical parameter (scalar) taking into account both ice crystals shape and size variability. The different values of asymmetry factor tested 5 here are obtained on the basis of nine pre-calculated phase functions for different ice crystal 6 shape (aggregate, plate, column, bullet and dendrite) and effective diameters comprise 7 between 4.5 and 150 μm that are used for cirrus clouds modelling [44]. The CIRCLE-2 phase 8 function is included as well. The asymmetry factor values obtained are between 0.70 and 9 100.85. Errors caused by undetected cirrus clouds versus asymmetry factors are displayed in 11Fig. 7. It appears that the underestimation decreases with the asymmetry factor. This is 12explained by the fact that cirrus cloud particles with a large value of g reflect less solar 13radiation than ice crystals with smaller asymmetry factor. Then, for a large value of *g*, more 14 solar radiation exhibited by the cirrus cloud will interact with tropospheric $NO₂$ located below 15the cloud. From the summary given in Table 3, error differences between $g = 0.70$ and $g =$ 160.85 are comprise between 0.07 % (for $c = 0.05$, $\tau = 0.05$ and low polluted condition) and 1721.3 % (for $c = 1$, $\tau = 3$ and heavily polluted condition). These results imply that, for fixed *c* 18 and τ , the error difference caused by undetected cirrus clouds with g lying between 0.70 and 190.85 (Table 3) is more important than the one caused by undetected cirrus clouds with *z* 20 comprise between 8 and 15 km (Table 1) or *h* comprise between 0.1 km and 3 km (Table 2). 21 The asymmetry factor appears to be a major source of error in tropospheric $NO₂$ retrieval in 22the presence of cirrus. 2

23

This sensitivity study showed that four cirrus parameters influence the retrieval of 25tropospheric $NO₂$ columns. These parameters are, in descending order, the cloud fraction c , 24

1the cloud optical depth τ , the asymmetry factor g and the cloud top height z . Since the error 2difference caused by cloud geometrical depth variability remains less than 3%, we have 3considered that cloud geometrical depth does not influence significantly the retrieval of 4tropospheric $NO₂$ columns. The $V_{t,ret}$ errors have also been estimated for other solar zenith 5angles (15 $^{\circ}$, 45 $^{\circ}$ and 60 $^{\circ}$) and other surface albedos (0.03 and 0.1) but they do not show large 6discrepancies with the results obtained for a solar zenith angle of 30° and a surface albedo of 0.05. 7

- 8
- 9
- 10
-
- 11
- 12

4. Error on NO2 retrieval caused by uncertainties in cirrus clouds properties 14 134.

15

The next step of our study is to look at the $V_{tr,ret}$ error induced by uncertainties of the cirrus 17 parameters that were identified previously as critical. The objective here is to evaluate the 18 precision needed for these parameters in order to constrain the retrievals. To this respect, we 19 simulated a true R_{TOA} using Eq. (3) with a set of cirrus properties (τ , g , z , h), a cloud fraction c 20 and a true tropospheric NO_2 vertical column $V_{tr,true}$. Then, $V_{tr,ret}$ is obtained by using as the 21input reflectance spectrum for the inversion algorithm, the clear-sky part of the TOA 22 reflectance R_{clear} . R_{clear} is a function of $R_{TOA}(p)$ and $R_{cloud}(p+\delta p)$ where p is τ , g, z, h or c . The 23 formulation of R_{clear} is presented for each cloud parameters in the following subsections. 16

1 Results are presented for moderately polluted condition as results found for the three polluted 2conditions did not differ significantly.

3

4.1 Cloud fraction 4

The error on tropospheric $NO₂$ column retrievals caused by uncertainty in cloud fraction is 6calculated as Eq. (4). The input reflectance spectrum of the inversion scheme is defined as: 5 7

8
$$
R_{clear} = \frac{R_{TOA}(\tau, g, z, h, c) - (c + \delta c)R_{cloud}(\tau, g, z, h)}{1 - (c + \delta c)},
$$
 (5)

9

10where δc is the uncertainty on cloud fraction. The simulations were performed for a cirrus 11cloud characterized by an asymmetry factor of 0.75, a cloud top height of 10 km and a 12 geometrical thickness of 1 km. The surface albedo, solar zenith angle and the viewing angle 13are the same as in Section 3. Fig. 8 displays errors on $V_{tr,ret}$ versus cloud optical depth for $\delta c =$ 140.01 (Fig. 8a), $\delta c = 0.05$ (Fig. 8b) and $\delta c = 0.09$ (Fig. 8c) and for three values of *c* (0.1, 0.5 15 and 0.8) representing low, medium and high cloud fractions. Overall, the overestimation of 16 cloud fraction results in an overestimation of tropospheric $NO₂$ columns (i.e., positive errors). 17If $c \leq 0.5$ (represented by lines with triangles or diamonds), overestimations are expected to 18be less than 10 % if $\delta c \le 0.05$ (Figs. 8a and Fig. 8b) and less than 20 % if $\delta c = 0.09$ (Fig. 8c). 19If $c \ge 0.8$, overestimations are expected to be higher, ranging from 10% to more than 100%, if $20\delta c \geq 0.05$ (Figs. 8b and 8c) for large cloud optical depth $(\tau > 1)$.

Within the framework of combining measurements performed by A-train sensors to 22improve OMI $NO₂$ products, cloud fraction can be derived from independent measurements. 23For example, MODIS can provide a subpixel cloud fraction information with a 1×1 km² 24 resolution. Accordingly, the cloud fraction estimation from MODIS within an OMI pixel is 25 expected to be achieved with lower uncertainty than the value of 0.05 reported by 21

1Koelemeijer et al. [45]. However, the estimation of the uncertainty on MODIS cloud fraction 2is a difficult task because it depends on the cloud type. Koren et al. [46] reported that for small cumulus cloud, MODIS cloud fraction is almost twice as the cloud fraction obtained 3 4from a finer spatial resolution (30 m) instrument. Fortunately, cirrus clouds have larger spatial 5extension than small cumulus clouds and would be less influenced by the scale dependence. Finally, these results highlight that cloud fraction within an OMI pixel should be determined 6 7with uncertainty lower than 0.05 in order to reduce the tropospheric $NO₂$ vertical columns 8errors, especially if the cloud optical depth is greater than 1.

9

4.2 Cloud optical depth 10

The error on tropospheric $NO₂$ columns retrieval caused by uncertainties in the cloud 12 optical depth is calculated as Eq. (4) where the retrieval is applied on the clear-sky part of the 13reflectance given by 11

14

15
$$
R_{clear} = \frac{R_{TOA}(\tau, g, z, h, c) - cR_{cloud}(\tau + \delta \tau, g, z, h)}{1 - c},
$$
 (6)

16

 17 where $\delta\tau$ is the uncertainty in cloud optical depth. The simulations were performed for a cirrus 18 cloud characterized by $g = 0.75$, $z = 10$ km and $h = 1$ km. The panel of Fig. 9 displays errors 19 on *V_{tr}* versus cloud optical depth for $\delta t/\tau$ = 10 % (Fig. 9a), $\delta \tau/\tau$ = 20 % (Fig. 9b) and $\delta \tau/\tau$ = 50 20% (Fig. 9c) and for three values of c (0.1, 0.5 and 0.8, differentiated by symbols). Here again, 21 the overestimation of cloud optical depth results in an overestimation of tropospheric $NO₂$ 22 columns. This is explained by the fact that an overestimation of τ will increase the cloud 23 reflectance. The numerator of Eq. (6) will then decrease leading to an overestimation of 24tropospheric NO₂ columns. For $c \le 0.5$, overestimations are expected to be less than 15 % if

δτ/τ ≤ 20 % (Figs. 9a and 9b). The same result is seen if *δτ/τ* = 50 % and *τ <* 1 (Fig. 9c). If *c* = 1 20.8, overestimations are expected to increase quickly, especially if $\delta \tau / \tau$ = 50 % (Fig. 9c).

By comparison, cirrus optical depth from MODIS and ground-based measurement has 4shown that MODIS overestimates the optical depth by 30% [47]. However, clouds with low 5optical depth (less than 0.4) are not retrieved from MODIS [48]. The use of CALIOP data will be helpful for a better characterisation of the optical depth since CALIOP is able to detect 6 7 clouds with τ < 0.05 [49]. Nevertheless, the use of CALIOP information is limited by the fact 8that CALIOP only overlaps with the nadir pixel of OMI. If no CALIOP data is available and 9the cirrus cloud is undetected by MODIS (*ie*, τ < 0.4), tropospheric NO₂ vertical column 10 errors are expected to be less than 10% (Fig. 4). 3

11

12

13

4.3 Asymmetry factor 14

15The error on tropospheric $NO₂$ columns retrieval caused by uncertainties in asymmetry factor 16is retrieved from the clear-sky part of the reflectance given by Eq. (6) and where the 17 uncertainty is applied to *g* instead of *τ*. Here again, g was chosen instead of the phase 18 function because it is a convenient integrated parameter function of the ice particle shape and 19effective size. So it is a first guess indicator of the general scattering behaviour of the cirrus 20 cloud. Fig. 10 displays errors versus cloud optical depth. R_{TOA} is simulated with $g = 0.75$, $z =$ 2110 km and $h = 1$ km. Here, the overestimation of asymmetry factor results in an 22underestimation of tropospheric $NO₂$ columns (i.e., negative errors). Underestimations up to 2340 % are expected if $c = 0.5$ (full lines) and up to 60 % if $c = 0.8$ (dashed lines), but do not 24 change that much whether $\delta g = 0.05$ (diamonds) or $\delta g = 0.1$ (triangles). These results show 25 that the uncertainty of the asymmetry factor is a major source of error in tropospheric $NO₂$

retrieval in the presence of cirrus. However, although modelling studies show that for ice 1 2crystals values of *g* could vary from 0.7 to 0.9 (see for instance the review paper of Baran 3[2009]), measurements do not show such variability in ice clouds. An uncertainty less than 40.05 can be expected since the asymmetry factor ranges more likely from 0.75 to 0.80 for ice 5clouds and from 0.80 to 0.85 for mixed phase clouds depending on the liquid water fraction 6[50,51]. However, the operational algorithms of MODIS retrieve the ice crystals effective 7radius (R_{eff}) and not g. Both information on R_{eff} and shape are needed to model the scattering 8properties of ice crystals. Usually, a combination of particles with different shapes and sizes is 9used as an equivalent microphysical model in MODIS retrieval process of the effective radius. 10The retrieved R_{eff} from MODIS corresponding to a specific microphysical shape model could 11be used to assess appropriate phase function and asymmetry parameter on the basis of 12MODIS Look-Up-Tables (LUT) [52,53]. However, MODIS R_{eff} seems to be overestimated by 13 comparison with Lidar retrievals [54].

4.4 Cloud top height 14

The error on tropospheric $NO₂$ column retrievals caused by uncertainties in the cloud top 16height is assumed from the clear-sky part of the reflectance given by Eq. (6) and where the 17 uncertainty is applied on *z* instead of τ . The panel of Fig. 11 displays errors on V_t versus 18 cloud optical depth for $\delta z = 0.5$ km (Fig. 11a), $\delta z = 1$ km (Fig. 11b) and $\delta z = 3$ km (Fig. 11c) 19 and for two values of *c* (0.5 and 0.8) and three values of *z* (8, 10 and 12 km). If $\delta z \le 1$ km 20(Figs. 11a and 11b), errors are expected to be less than 20 %. If $\delta z = 3$ km (Fig. 11c), errors 21are expected to be less than 20 % if $c \le 0.5$. Uncertainty of *z* retrieved from MODIS is 22 currently under investigation but preliminary results by comparison with Lidar measurements 23show difference on the order of 1 km [54]. However, this comparison remains limited due to 24the high occurrence of multi-layers ice clouds where MODIS only retrieves an effective cloud 15

top height. The use of CALIOP retrieval will allow constraining the cloud top height input in 1 2 order to retrieve the tropospheric $NO₂$ column in presence of cirrus cloud.

- 3
- 4

5. Contribution of subpixel cloud optical depth $55.$

6

The IPA formulation of the TOA reflectance in presence of clouds given by Eq. (3) does 8not consider the subpixel inhomogeneity of cloud properties. To this respect, we can consider 9that the optical thickness can be derived more accurately from high spatial resolution 10 measurements $(1\times1 \text{ km}^2)$ of MODIS. This information will be used to describe the variability 11of the optical thickness within an OMI pixel $(13\times24 \text{ km}^2)$. Therefore, two ways of 12 determining the contribution of the cloud to the total reflectance measured by OMI can be 13 considered: (1) by calculating R_{cloud} from the mean cloud optical depth $\langle t \rangle$ or (2) by 14 calculating the mean cirrus cloud reflectance $\langle R_{cloud} \rangle$ corresponding to a distribution of 15 optical depth within an OMI pixel. These two ways of calculating the contribution of the 16 cloud would have an important impact on the computing time of the retrieval because in the 17 first way, there is only one calculation for the mean cloud optical depth whereas in the second 18way there are several calculations for each value of the subpixel cloud optical depth 19 distributions. To compare these two ways, we have simulated different distribution of cloud 20optical depth considering a gamma distribution. Nine distributions have been selected 21 randomly with three mean cloud optical depths ($\langle \tau \rangle$ = 0.5, 1 or 2) and three variances (v_{τ} = 220.1, 1 or 2). In these distributions, 312 values have been considered representing the 23 maximum total number of MODIS subpixels within an OMI pixel at nadir. Cloud fractions 24 have been included in the study by considering 31 subpixels for $c = 0.1$, 62 subpixels for $c = 2$ 250.2, etc... to 281 subpixels for $c = 0.9$. For the cloud simulations we have used $z = 10$ km, $h =$ 7

11 km and $g = 0.75$. The influence of the solar zenith angle has been tested by considering two 2values, $\theta_s = 30^{\circ}$ and $\theta_s = 60^{\circ}$. Figure 12 illustrates the tropospheric NO₂ column error 3retrieved from the clear-sky part of the reflectance given by

4

$$
R_{clear} = \frac{\langle R_{TOA} \rangle - cR_{cloud}(\langle r \rangle, g, z, h)}{1 - c}, \quad (7)
$$

6

7where $\langle R_{TOA} \rangle$ is the mean TOA reflectance calculated as the average of TOA reflectances 8calculated using Eq. (3) for the different values of the cloud optical depth distribution. The 9tropospheric $NO₂$ column error resulting from the use of a mean cloud optical depth is either 10 positive or negative. The error remains negative for $\langle \tau \rangle = 0.5$ (dotted-dashed lines) and 11 remains positive for $\langle \tau \rangle = 2$ (dashed lines). Interestingly, the error is negative for $\langle \tau \rangle = 1$ 12(full lines) if $\theta_s = 30^\circ$ and positive if $\theta_s = 60^\circ$ showing the importance of the solar zenith angle 13in the subpixel inhomogeneity consideration. If $\theta_s = 30^\circ$ or 60°, errors caused by the use of a 14 mean cloud optical depth are expected to be less than \pm 15% for $c \le 0.5$. For cloud fractions 15greater than 0.5, errors increase quickly with cloud fraction and strongly with variances and 16cloud optical depth. The increase of errors with cloud fraction is explained by the 17 denominator 1-c in Eq. (7). The blow-up of errors at $SZA=60^\circ$ is explained by the bias 18between R_{cloud} calculated with the distribution of subpixels cloud optical depth and calculated 19 with the mean cloud optical depth. This bias increases with the variability of cloud optical 20 depths and solar zenith angle [55].

21 Finally, when the cloud fraction is lower than 0.5, the subpixel variability of the cloud optical 22depth does not need to be included in the cloud correction scheme. Otherwise, the subpixel 23 variability of the cloud optical depth is significant and should be considered in the cloud 24 correction scheme. Such information can be deduced from MODIS observations.

6. Conclusion $26.$

3

In this study, we have theoretically demonstrated the sensitivity of tropospheric $NO₂$ 5 $\text{column } (V_{tr})$ retrieval to the presence of cirrus clouds. In order of importance, we concluded 6that *c* (cloud fraction), τ (cloud optical depth), g (asymmetry factor of ice crystal phase 7 function) and *z* (cloud top height) influence the retrieval of V_{tr} . It appears from our 8simulations that V_{tr} is less influenced by h (cloud geometrical depth). The error on the 9 retrieved tropospheric $NO₂$ column depends also on the $NO₂$ profile. Because tropospheric $10NO₂$ is generally situated underneath the cirrus cloud, the shielding effect leads to 11 underestimating the tropospheric $NO₂$ columns for retrievals that do not correct for clouds in 12the first place. The underestimation depends strongly on c and τ ranging from few percents 13 for low cloud fraction to 55 % for $c = 1$ in heavily polluted conditions (i.e. for $V_{tr} = 9.36 \; 10^{15}$ 14molec/cm²). In the context of the A-Train constellation, the tropospheric $NO₂$ column in 15 presence of cirrus clouds can be retrieved by using the independent pixel approximation. The 16MODIS sensor coupled with CALIOP should be able to describe the optical and physical 17 properties of the clouds in such way to reduce the error on the tropospheric $NO₂$ column. The 18use of CALIOP information is however limited by the fact that CALIOP only overlaps with 19the nadir pixel of OMI. Nevertheless, errors on cloud properties will end up by errors on 20tropospheric $NO₂$ column retrieval. From our simulations, we have demonstrated that the error 21 will be reduced significantly if $c \leq 0.5$ as opposed to a retrieval that does not correct for 22 cirrus. In the case of a cirrus cloud correction based on IPA, c and τ would have to be known 23within accuracy better than 0.05 and 50 %, respectively. *z* would have to be known within 24 accuracy of at least 1 km and *g* would have to be be known within accuracy better than 0.05. 25The latter result shows that the uncertainty of the asymmetry factor is a major source of error 4

 1 in tropospheric $NO₂$ retrieval in the presence of cirrus. Using the subpixel information from 2MODIS cloud flag for c , from MODIS and CALIOP cloud products for τ and z and from 3MODIS R_{eff} for the phase function should be sufficient to reduce the error on tropospheric $4NO₂$ column retrieval. Under this circumstance, the use of a mean value of the cloud optical 5depth from subpixel information as provided by MODIS should be enough to take into 6account the cirrus cloud and speed up the retrieval significantly. If the cloud fraction is greater 7than 0.5, the subpixel variability of the cloud optical depth has to be included in the cloud 8correction scheme and can be obtained from MODIS. Moreover, differences between 9radiative transfer models used for cloud retrievals and sensor's calibration issues would 10introduce some discrepancies in the cloud correction scheme. The results in this study hold 11for retrievals that do not attempt to correct for cirrus clouds. Current state-of-science 12 retrievals do correct for clouds that include a range of situations including cirrus clouds. 13Nevertheless, the current $NO₂$ algorithm have not be evaluated in presence of cirrus. To 14interpret our results in the context of current retrieval algorithms, the sensitivity for cirrus 15 clouds in the O_2-O_2 (OMI) and O_2 -A (GOME(-2), SCIAMACHY) band based cloud retrievals 16needs to be studied. Such studies could point out whether a correction for cirrus based on 17 concurrent A-Train cloud information would be useful

18

19

Acknowledgments 20

The authors are grateful to Céline Cornet, Philippe Dubuisson and Nicolas Ferlay from 22the Laboratoire d'Optique Atmosphérique (LOA) for fruitful discussions. Authors are also 23grateful to Christophe Gourbeyre, Guillaume Mioche and Jean-François Gayet from the 24Laboratoire de Météorologie Physique (LaMP) for providing data from the CIRCLE-2 21

1

1campaign. This research was supported by the Centre National d'Etudes Spatiales (CNES) 2and the Institut National des Sciences de l'Univers (INSU/PNTS).

3

4

References 5

- $6[1]$ Crutzen PJ. The Role of NO and NO₂ in the Chemistry of the Troposphere and Stratosphere. Ann Rev Earth Planet Sci 1979;7:443-472. 7
- [2] Solomon S, Portmann RW, Sanders RW, Daniel JS, Madsen W, Bartram B, Dutton EG. 8
- On the role of nitrogen dioxide in the absorption of solar radiation. J Geophys Res 1999;104; D1012,047. 9 10
- 11[3] Burrows JP, Weber M, Buchwitz M, Rozanov V, Ladstätter-Weißenmayer A, Richter A,
- DeBeek R, Hoogen R, Bramstedt K, Eichmann KU, Eisinger M, Perner D. The Global 12
- Ozone Monitoring Experiment (GOME): Mission Concept and First Scientific Results. J 13
- Atm Sci 1999;56:151–175. 14

15[4] Bovensmann H, Burrows JP, Buchwitz M, Frerick J, Noël S, Rozanov VV, Chance K,

Goede PH. SCIAMACHY: Mission Objectives and Measurement Modes J Atm Sci 1999;56: 127–150. 16 17

18[5] Levelt PF, Hilsenrath E, Leppelmeier GW, Van den Oord GHJ, Bhartia PK, Tamminen J,

de Haan JF, Veefkind JP. Science objectives of the ozone monitoring instrument. IEEE 19

Trans Geosci Remote Sensing 2006;44:1199- 1208. 20

21[6] Platt U, Stutz J. Differential Optical Absorption Spectroscopy (DOAS): Principles and Applications. Heidelberg, Germany: Springer, 2006. 22

23[7] Beirle S, Platt U, Wenig M, Wagner T. Weekly cycle of $NO₂$ by GOME measurements: a

signature of anthropogenic sources. Atmos Chem Phys 2003 ;3 :2225-2232. 24

[8] Richter A, Burrows JP, Nüβ H, Granier C, Niemeier U. Increase in tropospheric nitrogen 1 dioxide over China observed from space. Nature 2005;437:129 – 132. 2

[9] Martin RV, Sioris CE,Chance K, Ryerson TB, Bertram TH, Wooldridge PJ, Cohen RC, 3 Neuman JA,Swanson A, Flocke FM. Evaluation of space-based constraints on global nitrogen oxide emissions with regional aircraft measurements over and downwind of eastern North America. J Geophys Res 2006;111:D15308. 4 5 6

7[10] Boersma KF, Jacob DJ, Eskes HJ, Pinder RW, Wang J, van der A RJ. Intercomparison of SCIAMACHY and OMI tropospheric NO₂ columns: Observing the diurnal evolution of chemistry and emissions from space. J Geophys Res 2008;113:D16S26. 8 9

10[11] Celarier EA, Brinksma EJ, Gleason JF, Veefkind JP, Cede A, Herman JR, Ionov D, Goutail F, Pommereau JP, Lambert JC, van Roozendael M, Pinardi G, Wittrock F, Schönhardt A, Richter A, Ibrahim OW,Wagner T, Bojkov B, Mount G, Spinei E, Chen CM, Pongetti TJ, Sander SP, Bucsela EJ, Wenig MO, Swart DPJ, Volten H, Kroon M, Levelt PF. Validation of Ozone Monitoring Instrument nitrogen dioxide columns. J Geophys Res 2008;113: D15S15. 11 12 13 14 15

16[12] Wenig MO, Cede AM, Bucsela EJ, Celarier A, Boersma KF, Veefkind JP, Brinksma EJ, Gleason JF, Herman JR. Validation of OMI tropospheric $NO₂$ column densities using direct-Sun mode Brewer measurements at NASA Goddard Space Flight Center. J Geophys Res 2008;113:D16S45. 17 18 19

20[13] van Noije TPC, Eskes HJ, Dentener FJ, Stevenson DS, Ellingsen K, Schultz MG, Wild O, Amann M, Atherton CS, Bergmann DJ, Bey I, Boersma KF, Butler T, Cofala J, Drevet J, Fiore AM, Gauss M, Hauglustaine DA, Horowitz LW, Isaksen ISA, Krol MC, Lamarque JF, Lawrence MG, Martin RV, Montanaro V, Müller JF, Pitari G, Prather MJ, Pyle JA, Richter A, Rodriguez JM, Savage NH, Strahan SE, Sudo K, Szopa S, 21 22 23 24

1

van Roozendael M. Multi-model ensemble simulations of tropospheric $NO₂$ compared with GOME retrievals for the year 2000. Atmos Chem Phys 2006;6:2943-2979. 1 2

[14] Blond N, Boersma KF, Eskes HJ, van der A RJ, Van Roozendael M, De Smedt I, 3 Bergametti G, Vautard R. Intercomparison of SCIAMACHY nitrogen dioxide observations, in situ measurements and air quality modeling results over Western Europe. J Geophys Res 2007;112:D10311. 4 5 6

7[15] Bucsela EJ, Perring AE, Cohen RC, Boersma KF. Celarier A. Gleason F, Wenig M O,

Bertram TH, Wooldridge PJ, Dirksen R, Veefkind JP. Comparison of tropospheric NO₂ from in situ aircraft measurements with near-real-time and standard product data from OMI. J Geophys Res 2008;113:D16S31. 8 9 10

11[16] Boersma KF, Eskes HJ, Brinksma EJ. Error analysis for tropospheric $NO₂$ retrieval from space. J Geophys Res 2004;109:D04311. 12

13[17] Krijger JM, van Weele M, Aben I, Frey R. Technical Note: The effect of sensor resolution on the number of cloud-free observations from space. Atmos Chem Phys 2007;7:2881-2891. 14 15

16[18] Koelemeijer R, Stammes P. Effects of clouds on ozone column retrieval from GOME UV measurements. J Geophys Res 1999;104:8281–8294. 17

18[19] Liu X, Newchurch MJ, Loughman R, Bhartia PK. Errors resulting from assuming opaque Lambertian clouds in TOMS ozone retrieval. JQSRT 2004;85:337-365. 19

20[20] Kokhanovsky AA, Mayer B, Rozanov VV, Wapler K, Lamsal LN, Weber M, Burrows J

P, Schumann U. Satellite Ozone Retrieval Under Broken Cloud Conditions: An Error Analysis Based on Monte Carlo Simulations. IEEE Trans Geosci Remote Sensing 2007;45:187-194. 21 22 23

24^[21] Kokhanovsky AA, Rozanov VV. The uncertainties of satellite DOAS total ozone retrieval for a cloudy sky. Atmos Res 2008;87: 27-36. 25

1

1[22] Stammes P, Sneep M, de Haan JF, Veefkind JP, Wang P, Levelt PF. Effective cloud fractions from the Ozone Monitoring Instrument: Theoretical framework and validation. J Geophys Res 2008;113:D16S38. 2 3

4[23] Wang P, Stammes P, Boersma KF. Impact of Effective Cloud Fraction Assumption on Tropospheric NO2 Retrievals. Proc of the Atmospheric Science Conference, 8-12 May 2006 at ESRIN, Frascati Italy. Edited by H. Lacoste and L. Ouwehand. European Space Agency, 2006;SP-628:72.1. 5 6 7

[24] Acarreta JR, De Haan JF, Stammes P. Cloud pressure retrieval using the O2-O2 8 absorption band at 477 nm. J Geophys Res 2004;109:D05204. 9

10[25] van Diedenhoven B, Hasekamp OP, Landgraf J. Retrieval of cloud parameters from satellite-based reflectance measurements in the ultraviolet and the oxygen A-band. J Geophys Res 2007;112:D15208. 11 12

13[26] Platnick S, King MD, Ackerman SA, Menzel WP, Baum BA, Riedi JC, Frey RA. The 14MODIS cloud products: algorithms and examples from Terra. IEEE Trans Geosci Remote 15Sensing 2003;41:459- 473.

16[27] Wylie DP, Menzel WP, Woolf HM, Strabala KI. Four Years of Global Cirrus Cloud Statistics using HIRS. J Clim1994;7(12):1972-1986,. 17

18[28] Nazaryan H, McCormick MP, Menzel WP. Global characterization of cirrus clouds using

CALIPSO data. J Geophys Res 2008;113:D16211. 19

20[29] Rozanov AA, Rozanov VV, Buchwitz M, Kokhanovsky AA, Burrows JP. SCIATRAN 2.0-a new radiative transfer model for geophysical applications in the 175-2400 nm spectral range. Adv Space Res 2005;36:1015-1019. 21 22

[30] Boersma KF, Eskes HJ, Veefkind JP, Brinksma EJ, van der A RJ, Sneep M, van den Oord GHJ, Levelt PF, Stammes P, Gleason JF, Bucsela EJ. Near-real time retrieval of tropospheric $NO₂$ from OMI. Atmos Chem Phys 2007;7:2103-2118. 2301 24 25

1

[31] Pfeilsticker K, Erle F, Funk O, Marquard L, Wagner T, Platt U. Optical path 1 modifications due to tropospheric clouds: Implications for zenith sky measurements of stratospheric gases. J Geophys Res 1998;103(D19):25323–25335. 2 3

[32] Oshchepkov S, Isaka, H, Gayet JF, Sinyuk A, Auriol F, Havemann S. Microphysical 4 properties of mixed-phase and ice clouds retrieved from in situ airborne "Polar Nephelometer" measurements, Geophys. Res. Lett., 27, 209-213, 2000 5 6

7[33] Jourdan O, Oshchepkov S, Shcherbakov V, Gayet JF, Isaka H. Assessment of cloud optical parameters in the solar region: Retrievals from airborne measurements of scattering phase functions. J Geophys Res 2003;108:D134572. 8 9

10[34] Yang P, Liou KN, Geometric-optics-integral-equation method for light scattering by nonspherical ice crystals. Appl Opt 1996;35:6568-6584. 11

12[35] Shcherbakov VN, Gayet JF, Jourdan O, Minikin A, Ström J, Petzold A. Assessment of Cirrus Cloud Optical and Microphysical Data Reliability by Applying Statistical Procedures. J Atmos Ocean Technol 2005;22:409-420. 13 14

15[36] Gayet JF, Shcherbakov VN, Mannstein H, Minikin A, Schumann U, Ström J, Petzold A Ovarlez J, Immler F. Microphysical and optical properties of midlatitude cirrus clouds observed in the southern hemisphere during INCA. QJR Meteor Soc 2006;132;2719-2748. 18[37] Francis PN, Foot JS, Baran AJ. Aircraft measurements of the solar and infrared radiative properties of cirrus and their dependence on ice crystal shape. J Geophys Res 1999;104:31685-31696. 16 17 19 20

21[38] Labonnote LC, Brogniez G, Buriez JC, Doutriaux-Boucher M, Gayet JF, Macke A. Polarized light scattering by inhomogeneous hexagonal monocrystals : Validation with ADEOS-POLDER measurements. J Geophys Res 2001;106:12139-12154. 22 23

1

[39] Baran AJ, Francis, PN. On the radiative properties of cirrus cloud at solar and thermal 1 wavelengths: A test of model consistency using high-resolution airborne radiance measurements. QJR Meterorol Soc, 2004;130;763-778. 2 3

4[40] Baran AJ, Labonnote LC. On the reflection and polarisation properties of ice cloud. JQSRT, 2006;100:41-54. 5

6[41] Baran AJ, Labonnote LC. A self-consistent scattering model for cirrus. I: The solar region. QJR Meterorol Soc 2007;133:1899-1912. 7

 $8[42]$ Dufour E and Bréon FM. Spaceborne Estimate of Atmospheric $CO₂$ Column by Use of the Differential Absorption Method: Error Analysis. Appl Opt 2003;42:3595-3609. 9

10[43] Schneising O, Buchwitz M, Burrows JP, Bovensmann H, Reuter M, Notholt J, Macatangay R, Warneke T. Three years of greenhouse gas column-averaged dry air mole fractions retrieved from satellite – Part 1: Carbon dioxide. Atmos Chem Phys 2008;8:3827-3853. 11 12 13

14[44] Yang P, Liou KN. Single-scattering properties of complex ice crystals in terrestrial atmosphere. Contr Atmos Phy 1998;71:223–248. 15

16 [45] Koelemeijer R, Stammes P, Hovenier J, de Haan J. A fast method for retrieval of cloud parameters using oxygen A band measurements from the Global Ozone Monitoring Experiment. J Geophys Res 2001;106(D4):3475-3490. 17 18

19[46] Koren I, Oreopoulos L, Feingold G, Remer LA, Altaratz O. How small is a small cloud? Atmos Chem Phys 2008;8:3855-3864. 20

21[47] Mace GG, Zhang Y, Platnick S, King MD, Minnis P, Yang P. Evaluation of Cirrus Cloud Properties Derived from MODIS Data Using Cloud Properties Derived from Ground-Based Observations Collected at the ARM SGP Site. J Appl Meteor 2005;44:221–240. 24[48] Ackerman SA, Holz RE, Frey R, Eloranta EW, Maddux BC, McGill M. Cloud Detection 22 23

with MODIS. Part II: Validation. J Atmos Oceanic Technol 2008;25:1073–1086. 25

1

1[49] Winker DM, Pelon JR, McCormick MP. The CALIPSO mission: spaceborne lidar for observation of aerosols and clouds. Proc SPIE 2003;4893:1-11. 2

3[50] Garrett TJ, Hobbs P., Gerber H. Shortwave, single-scattering properties of arctic ice clouds. J Geophys Res 2001:106(D14);15155-15172. 4

[51] Gayet JF, Asano S, Yamazaki A, Uchiyama A, Sinyuk A, Jourdan O, Auriol F. Two 5 case studies of winter continental-type water and mixed-phase stractocumuli over the sea, Part I : Microphysical and optical properties. J Geophys Res 2002:107(D21);doi:10.1029/2001JD001106. 6 7 8

9[52] Baum BA, Heymsfield AJ, Yang P, Bedka ST. Bulk scattering models for the remote sensing of ice clouds. Part 1: Microphysical data and models. J Appl Meteor 2005;44 :1885-1895. 10 11

12[53] Baum BA, Yang P, Heymsfield AJ, Platnick S, King MD, Hu YX, Bedka ST. Bulk scattering models for the remote sensing of ice clouds. Part 2: Narrowband models. J Appl Meteor 2005;44:1896-1911. 13 14

15[54] Chiriaco M, Chepfer H, Minnis P, Haeffelin M, Platnick S, Baumgardner D, Dubuisson P, McGill M, Noël V, Pelon J, Spangenberg D, Sun-Mack S, Wind G. Comparison of CALIPSO-Like, LaRC, and MODIS Retrievals of Ice-Cloud Properties over SIRTA in France and Florida during CRYSTAL-FACE. J Appl Meteor Clim 2007;46:249-272. 16 17 18 19

20[55] Cahalan RF, Ridgway W, Wiscombe WJ, Bell TL, Snider JB. The Albedo of Fractal Stratocumulus Clouds. J Atmos Sci 1994:51;2434–2455. 21

22Baran AJ. A review of the light scattering properties of cirrus, 2009 JQSRT 2009;110:1239-1260. 23

1

1Kleipool Q. L., M. R. Dobber, J. F. de Haan, P. F. Levelt (2008), Earth surface reflectance

- climatology from 3 years of OMI data, J. Geophys. Res., 113, D18308, 2
- doi:10.1029/2008JD010290. 3

Boersma, K. F., Dirksen, R., Veefkind, J. P., Eskes, H. J., and 4

5van der A, R. J.: Dutch OMI NO2 (DOMINO) data product

6HE5 data file user manual, TEMIS website, http://www.temis.

nl/airpollution/no2.html, 2008b. 7

Figure captions 1

2Figure 1. Atmospheric profiles of $NO₂$ Volume Mixing Ratio in ppbv used in the study.

3Figure 2. Cirrus cloud phase function inferred from aircraft measurement during the CIRCLE2 campaign. 4

Figure 3. Differential optical densities for heavily polluted troposphere without cloud (green 5 6line), with $c = 1$ (blue line) or $c = 0.5$ (red line).

 7 Figure 4. NO₂ tropospheric error caused by undetected cirrus cloud as function of cloud 8 optical depth and cloud fraction. Here, $z = 10$ km, $h = 1$ km and $g = 0.75$.

 9 Figure 5. NO₂ tropospheric error caused by undetected cirrus cloud versus cloud top height. 10Green, bleu and red lines stand for low, moderately and heavily polluted conditions. Cloud 11 fraction and cloud optical depth were assumed to be 0.5 and 1, respectively for diamonds and 12to be 1 and 3 for triangles. Here, $h = 1$ km and $g = 0.75$.

13 Figure 6. Same as Fig. 5 versus cloud geometrical depth. Here, $z = 10$ km and $q = 0.75$.

14 Figure 7. Same as Fig. 5 versus asymmetry factor. Here, $z = 10$ km and $h = 1$ km.

15 Figure 8. NO₂ tropospheric column error caused by an uncertainty in the cloud fraction $\delta c =$ 160.01 (a), $\delta c = 0.05$ (b) and $\delta c = 0.09$ (c) versus cloud optical depth and for different cloud 17 fraction ($c = 0.1$, 05 and 0.8, differentiated by symbols)., Here, $z = 10$ km, $h = 1$ km and $g =$ 180.75 and moderately polluted condition is assumed.

19 Figure 9. Same as Fig. 8 for $\delta \tau / \tau$ = 10% (a), $\delta \tau / \tau$ = 20% (b) and $\delta \tau / \tau$ = 50% (c).

20 Figure 10. NO₂ tropospheric column error caused by an uncertainty in the asymmetry factor $21\delta g = 0.05$ (diamonds) and $\delta g = 0.1$ (triangles) versus cloud optical depth for $c = 0.5$ (full lines) 22and $c = 0.8$ (dashed lines). Here, $z = 10$ km, $h = 1$ km and $g = 0.75$ and moderate polluted 23 condition is assumed.

24 Figure 11. NO₂ tropospheric column error caused by an uncertainty in the cloud top height δ*z* $25= 0.5$ km (a), $\delta z = 1$ km (b) and $\delta z = 3$ km (c) versus cloud optical depth and for c = 0.5 (full 1lines), $c = 0.8$ (dashed lines), $z = 8$ km (diamonds) and $z = 12$ km (triangles). Here, $h = 1$ km 2 and $g = 0.75$ and moderate polluted condition is assumed.

3Figure 12. $NO₂$ tropospheric column error caused by the use of a mean cloud optical depth 4 instead of subpixels information versus cloud fraction if $\theta_s = 30^\circ$ (left) and if $\theta_s = 60^\circ$ (right).

Table captions

2Table 1. NO₂ tropospheric error difference caused by cloud top height *z* variability. Values in 3each box of fixed c and τ correspond to low, moderately and heavily polluted conditions, 4respectively.

Table 2. Same as Table 1 for cloud geometrical depth *h*.

Table 3. Same as Table 1 for asymmetry factor *g*.

2Figure 1. Atmospheric profiles of $NO₂$ Volume Mixing Ratio in ppbv used in the study.

1

2Figure 2. Cirrus cloud phase function inferred from aircraft measurement during the 3CIRCLE2 campaign.

2Figure 3. Differential optical densities for heavily polluted troposphere without cloud (green 3line), with $c = 1$ (blue line) or $c = 0.5$ (red line).

Figure 4. NO₂ tropospheric error caused by undetected cirrus cloud as function of cloud 3 optical depth and cloud fraction. Here, $z = 10$ km, $h = 1$ km and $g = 0.75$.

Figure 5. NO₂ tropospheric error caused by undetected cirrus cloud versus cloud top height. 3Green, bleu and red lines stand for low, moderately and heavily polluted conditions. Cloud fraction and cloud optical depth were assumed to be 0.5 and 1, respectively for diamonds and 5to be 1 and 3 for triangles. Here, $h = 1$ km and $g = 0.75$.

2 Figure 6. Same as Fig. 5 versus cloud geometrical depth. Here, $z = 10$ km and $q = 0.75$.

2 Figure 7. Same as Fig. 5 versus asymmetry factor. Here, $z = 10$ km and $h = 1$ km.

2Figure 8. NO₂ tropospheric column error caused by an uncertainty in the cloud fraction $\delta c =$ 30.01 (a), $\delta c = 0.05$ (b) and $\delta c = 0.09$ (c) versus cloud optical depth and for different cloud 4 fraction ($c = 0.1$, 05 and 0.8, differentiated by symbols)., Here, $z = 10$ km, $h = 1$ km and $g =$ 50.75 and moderately polluted condition is assumed.

2 Figure 9. Same as Fig. 8 for $\delta \tau / \tau$ = 10% (a), $\delta \tau / \tau$ = 20% (b) and $\delta \tau / \tau$ = 50% (c).

 2 Figure 10. NO₂ tropospheric column error caused by an uncertainty in the asymmetry factor $3\delta g = 0.05$ (diamonds) and $\delta g = 0.1$ (triangles) versus cloud optical depth for $c = 0.5$ (full lines) 4and $c = 0.8$ (dashed lines). Here, $z = 10$ km, $h = 1$ km and $g = 0.75$ and moderate polluted 5condition is assumed.

2Figure 11. NO₂ tropospheric column error caused by an uncertainty in the cloud top height δ*z* $3= 0.5$ km (a), $\delta z = 1$ km (b) and $\delta z = 3$ km (c) versus cloud optical depth and for c = 0.5 (full 4lines), $c = 0.8$ (dashed lines), $z = 8$ km (diamonds) and $z = 12$ km (triangles). Here, $h = 1$ km 5 and $g = 0.75$ and moderate polluted condition is assumed.

Figure 12. NO₂ tropospheric column error caused by the use of a mean cloud optical depth 3instead of subpixels information versus cloud fraction if $θ_s = 30°$ (left) and if $θ_s = 60°$ (right).

2Table 1. NO₂ tropospheric error difference caused by cloud top height *z* variability. Values in 3each box of fixed c and τ correspond to low, moderately and heavily polluted conditions, 4respectively.

Table 2. Same as Table 1 for cloud geometrical depth *h*. 2

Table 3. Same as Table 1 for asymmetry factor *g*. 2