Vertical profiles of global tropospheric nitrogen dioxide (NO2) obtained by cloud-slicing TROPOMI

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- 10 **Abstract.** Routine observations of the vertical distribution of tropospheric nitrogen oxides ($NO_x \equiv NO + NO₂$) are severely lacking, despite the large influence of NO_x on climate, air quality, and atmospheric oxidants. Here we derive vertical profiles of global seasonal mean tropospheric NO2 by applying the cloud-slicing method to TROPOspheric Monitoring Instrument (TROPOMI) columns of NO₂ retrieved above optically thick clouds. The resultant NO₂ are at a horizontal resolution of $1^{\circ} \times$ 1° for multiple years (June 2018 to May 2022) covering 5 layers in the upper (180-320 hPa and 320-450 hPa) and mid (450-
- 15 600 hPa and 600-800 hPa) troposphere, and the marine boundary layer (800 hPa to the Earth's surface). Terrestrial boundary layer NO2 are obtained as the difference between TROPOMI tropospheric columns and the integrated column of cloud-sliced NO₂ in all layers above the boundary layer. Cloud-sliced NO₂ is typically 20-60 pptv throughout the free troposphere and spatial coverage ranges from $>60\%$ in the mid-troposphere to $< 20\%$ in the upper troposphere and boundary layer. Our product is similar (within 10-15 pptv) to NO₂ data from NASA DC-8 aircraft campaigns (INTEX-A, INTEX-B, ARCTAS, SEAC⁴RS,
- 20 ATom) when both datasets are abundant and sampling coverage is commensurate, but such instances are rare. We use the cloud-sliced NO2 to critique current knowledge of the vertical distribution of global NO2, as simulated with the GEOS-Chem chemical transport model updated to include peroxypropionyl nitrate (PPN) and aerosol nitrate photolysis that liberate NO₂ in the lower and mid-troposphere for aerosol nitrate photolysis and upper troposphere for PPN. Multiyear GEOS-Chem and cloud-sliced means are compared to mitigate the influence of interannual variability. We find that for cloud-sliced NO2 the
- 25 interannual variability is \sim 10 pptv over remote areas and \sim 25 pptv over areas influenced by lightning and surface sources. The model consistently underestimates $NO₂$ across the remote marine troposphere by \sim 15 pptv. In the northern midlatitudes, GEOS-Chem overestimates mid-tropospheric NO₂ by 20-50 pptv, as NO_x production per lightning flash is parameterised to be almost double the rest of the world. There is a critical need for in-situ NO₂ measurements in the tropical terrestrial troposphere to evaluate cloud-sliced NO2 there. The model and cloud-sliced NO2 discrepancies identified here need to be
- 30 investigated further to ensure confident use of models to understand and interpret factors affecting the global distribution of tropospheric NOx, ozone and other oxidants.

1 Introduction

In the troposphere, nitrogen oxides ($NO_x \equiv NO + NO₂$) influence the formation of tropospheric ozone (O₃), a greenhouse gas, and the hydroxyl radical (OH), the main atmospheric oxidant (Atkinson, 2000; Bloss et al., 2005). Due to its influence on OH,

- 35 NO_x also indirectly affects the lifetime and abundance of the potent greenhouse gas methane (Wild et al., 2001) and nonmethane volatile organic compounds that contribute to O₃ and particulate matter pollution (Crutzen and Andreae, 1990; Karl et al., 2007; Marais et al., 2016). NOx is directly emitted from high-temperature combustion of fossil fuels, from open and domestic burning of biomass, and from natural processes such as lightning and bacteria in soils (Dignon, 1992; Pickering et al., 1998; Jain et al., 2006; Vinken et al., 2014). NO_x also enters the upper layers of the troposphere via downwelling from the
- 40 stratosphere (Poulida et al., 1996). The distribution of NO_x varies throughout the troposphere as a result of these sources and due to recycling of NO_x via oxidation, photolysis and thermal decomposition of gas- and aerosol-phase reservoirs of nitrogen (Chatfield, 1994; Moxim et al., 1996; Kotamarthi et al., 2001; Scharko et al., 2014). In the warm lower troposphere where anthropogenic sources dominate, the lifetime of NO_x is a few hours. This increases with altitude to several days in the cold, dry upper troposphere where NO_x is present mostly as NO (Travis et al., 2016), reservoir compounds dominate, and terminal

45 loss of NO_x via wet deposition in the form of nitric acid (HNO₃) is limited (Jaeglé et al., 1998).

Knowledge of the vertical distribution of tropospheric NO_x has been largely informed by in-situ instruments on research and commercial aircraft (Crawford et al., 1996; Brenninkmeijer et al., 1999; Bradshaw et al., 2000; Emmons et al., 2000; Petzold et al., 2015; Stratmann et al., 2016). These aircraft campaigns are few in time and space. The instruments used to measure $NO₂$

- 50 are also susceptible to interference from decomposition of thermally unstable reservoir compounds of NO_x (Bradshaw et al., 2000; Browne et al., 2011; Reed et al., 2016). This interference is most severe in the upper troposphere and in remote marine regions where thermally labile NO_x reservoir compounds are abundant and decomposition of these compounds is promoted by the warm instrument inlet (Murphy et al., 2004; Nault et al., 2015; Shah et al., 2023). Studies now supplement these measurements with calculated daytime $NO₂$ concentrations, as NO and $NO₂$ can be assumed to be in photochemical steady
- 55 state (PSS) (Davis et al., 1993; Crawford et al., 1996).

Networks of ground-based remote sensing instruments such as Multi Axis Differential Optical Absorption Spectroscopy (MAX-DOAS) and direct-sun Pandora instruments have expanded globally. Still, geographic coverage for both is mostly in the northern hemisphere (Verhoelst et al., 2021). For Pandora, only the total tropospheric column can be derived from total 60 atmospheric column measurements (Pinardi et al., 2020). MAX-DOAS, under ideal conditions, can retrieve up to four independent layers in the troposphere, though vertical extent at most sites excludes the upper troposphere (Tirpitz et al., 2021). Space-based remote sensing observations used to retrieve vertical column densities (VCDs) of tropospheric NO₂ address

limited spatial sampling of commercial and research aircraft and Pandora and MAX-DOAS networks by offering daily global coverage, but with only one piece of vertical information in the troposphere (Ryan et al., 2023). These satellite observations

- 65 are also impacted by biases in modelled vertical profiles of NO2 required to retrieve VCDs (Verhoelst et al., 2021), in particular in the upper troposphere where satellite observations are most sensitive to tropospheric NO₂ (Boersma et al., 2004; Travis et al., 2016; Silvern et al., 2018; Shah et al., 2023).
- Mixing ratios of NO₂ in distinct layers of the troposphere can be retrieved using so-called cloud-slicing. This technique targets 70 partial columns (stratospheric + tropospheric) above clouds that are sufficiently optically thick that UV-visible instruments observe discrete layers in the troposphere. Cloud-slicing was first applied by Ziemke et al. (2001) to O_3 columns to derive seasonal multi-year mean upper tropospheric O_3 mixing ratios in the tropics. Cloud-slicing has since been used to retrieve seasonal mean concentrations of NO2 from the Ozone Monitoring Instrument (OMI) in both the mid (900-650 hPa or 2-4 km) and upper (450-280 hPa or 6-11 km) troposphere at 5° latitude $\times 8^{\circ}$ longitude (500 km $\times 800$ km) as well as in six pressure 75 levels (centred at 280, 380, 500, 620, 720 and 820 hPa) at $2^{\circ} \times 2^{\circ}$ (Choi et al., 2014; Belmonte Rivas et al., 2015; Marais et al., 2018). The OMI cloud-sliced NO₂ data provide useful information at very coarse scales ($20^{\circ} \times 32^{\circ}$, seasonal) (Marais et al., 2018) and are hindered by large data loss after 2007 when many satellite pixels became obscured by the row anomaly (Torres et al., 2018). More recently, the higher spatial resolution TROPOspheric Monitoring Instrument (TROPOMI) has been used to derive NO₂ mixing ratios in the upper troposphere (450-180 hPa or 6-12 km) at finer scales than was possible with 80 OMI of $1^{\circ} \times 1^{\circ}$ (~100 km) (Marais et al., 2021). Cloud-sliced NO₂ from TROPOMI has so far only been derived for a single year, as at the time there were frequent updates to the retrieval that led to inconsistencies in the TROPOMI NO2 VCDs used
- Evaluation of cloud-sliced NO2 data products is very limited, as coincidence of satellite observations and aircraft campaigns 85 is rare. Choi et al. (2014) found that the NASA OMI mid tropospheric product is similar (<10% difference) to coincident research aircraft campaign observations, limited to Texas and the Pacific Ocean west of North America. Marais et al. (2021) intercompared seasonal mean cloud-sliced upper tropospheric NO2 from TROPOMI and the NASA OMI product to identify that TROPOMI background values routinely exceed OMI by 12-26 pptv. Given these product disparities, independent evaluation of cloud-sliced NO2 mixing ratios is crucial. Past (2006-2013) NASA DC-8 aircraft campaigns and the more recent 90 (2016-2018) NASA DC-8 Atmospheric Tomography Mission (ATom) measurement campaign sampled the troposphere from close to the surface to the upper layers of the troposphere, offering the opportunity to evaluate cloud-sliced $NO₂$ mixing ratios over the remote Pacific and Atlantic Oceans (ATom) (Thompson et al., 2022), the Canadian Arctic during the Arctic Research of the Composition of the Troposphere from Aircraft and Satellites (ARCTAS) campaign (Jacob et al., 2010), the eastern US during the Intercontinental Chemical Transport Experiment – North America Phases A and B (INTEX-A and INTEX-B)

for cloud slicing. TROPOMI NO2 data have since been reprocessed to obtain a consistent data record starting in May 2018.

95 (Singh et al., 2006, 2009) and the Studies of Emissions and Atmospheric Composition, Clouds and Climate Coupling by Regional Surveys (SEAC⁴RS) (Toon et al., 2016) campaigns, and the northern Pacific during INTEX-B.

Here we derive a global dataset of 4 years of seasonal multiyear mean concentrations of NO₂ in five discrete vertical layers in the troposphere from the planetary boundary layer to the upper troposphere. We evaluate our dataset against directly measured

100 and calculated (PSS) NO2 from multiple NASA DC-8 aircraft campaigns and go on to use the cloud-sliced data to assess current understanding of the global vertical distribution of tropospheric NOx as simulated by the GEOS-Chem chemical transport model.

2 Methods

2.1 Cloud-slicing TROPOMI NO2 columns

- 105 TROPOMI was launched in October 2017 aboard the Sentinel-5P satellite. The initial TROPOMI nadir spatial resolution of 7.2 km \times 3.5 km was enhanced to 5.6 km \times 3.5 km in August 2019 (Liu et al., 2021). The swath width is 2600 km, resulting in daily global coverage at an equator crossing time of 13:30 local solar time (LST). To derive our cloud-sliced product, we use TROPOMI Level 2 swaths retrieved using a consistent algorithm (version 2.3.1). Data are available as the reprocessed (PAL) product from 1 June 2018 to 14 November 2021 (https://data-portal.s5p-pal.com/, last accessed 17th February 2022)
- 110 and as the offline (OFFL) product from 14 November 2021 to 31 May 2022 (https://s5phub.copernicus.eu/dhus/#/home, last accessed 7 July 2022; now available at https://dataspace.copernicus.eu/browser/). The cloud-slicing approach was first applied to TROPOMI by Marais et al. (2021) to derive NO2 mixing ratios in the upper troposphere over a broad pressure range from 450 to 180 hPa. We apply this cloud-slicing, with updates detailed below, to the whole troposphere to derive vertical profiles of seasonal mean NO₂ at the same $1^{\circ} \times 1^{\circ}$ resolution as Marais et al. (2021) for multiple years (2018-2022) over five pressure
- 115 ranges: one in the boundary layer below 800 hPa (\leq -2 km), two in the mid-troposphere at 800-600 hPa (\sim 2-4 km) and 600-450 hPa (~4-6 km), and two in the upper troposphere at 450-320 hPa (~6-9 km) and 320-180 hPa (~9-12 km).

The first application of cloud-slicing to TROPOMI NO2 is described in detail in Marais et al. (2021). We mostly follow the same approach. That is, pixels of individual swaths are filtered to isolate observations above optically thick clouds (cloud

- 120 radiance fraction > 0.7). These are binned into cloud-top pressures within the 5 targeted pressure ranges on a fixed $1^{\circ} \times 1^{\circ}$ grid. The stratospheric component of the total VCDs is corrected for a 13% underestimate in variance identified by Marais et al. (2021) from comparison to ground-based direct sun photometer Pandora measurements at the high-altitude (4.2 km) Mauna Loa site. The corrected stratospheric VCDs are multiplied by the reported stratospheric air mass factors (AMFs) to calculate stratospheric slant columns. The stratospheric slant columns are then subtracted from the total slant columns to estimate
- 125 tropospheric slant columns that are converted to tropospheric VCDs using a geometric AMF. Only clusters of total abovecloud VCDs with a relatively uniform stratosphere are retained for cloud-slicing. These are identified as clusters of $1^{\circ} \times 1^{\circ}$ pixels with a stratospheric column relative standard deviation < 0.02. A uniform stratosphere ensures that variability in partial NO₂ columns above optically thick clouds is dominated by variability in the troposphere. Cloud-slicing also requires that each

cluster include a representative range of cloud-top pressures (Choi et al., 2014). To ensure this is achieved, we remove clusters

- 130 with cloud pressure ranges that are < 60% of the pressure range of each layer (for example, 120 hPa threshold for the 800-600 hPa layer) and that have a large standard deviation ($>$ 30 hPa), which is consistent with cloud-slicing by Choi et al. (2014) and Marais et al. (2018, 2021).
- Next, we regress cloud-top pressures against above-cloud $NO₂ VCDs$ for clusters with at least 10 satellite pixels. We replace 135 the reduced major axis (RMA) regression fit originally used by Marais et al. (2021) with Theil regression, as this reduces influence from outliers and is better suited to data that are not always normally distributed (Theil, 1950; Sen, 1968). The regression slope in molecules cm⁻² hPa⁻¹ is converted to NO₂ volume mixing ratios in pptv as in Equation (5) of Choi et al. (2014). The updated Theil regression fit addresses the 12-26 pptv overestimate in background values of cloud-sliced upper tropospheric NO2 identified by Marais et al. (2021) from comparison to the OMI upper tropospheric product. It also negates 140 the need for the large TROPOMI free tropospheric column $NO₂$ bias correction that Marais et al. (2021) used to resolve an apparent overestimate in TROPOMI compared to free tropospheric NO2 columns derived with measurements from Pandora 145 approach to those from Marais et al. (2021). To ensure a consistent comparison, we recompute our updated cloud-sliced $NO₂$
- and MAX-DOAS instruments at the high-altitude Izaña site. We also find that the outlier filter used by Marais et al. (2021) for cloud-sliced $NO₂ > 200$ pptv is no longer needed, as it has negligible impact on seasonal mean cloud-sliced $NO₂$ using our updated approach. As an initial assessment, we compare upper tropospheric cloud-sliced NO2 from our updated cloud-sliced
- to cover the same pressure range (450-180 hPa) and time (June 2019 to May 2020) as Marais et al. (2021) and only compare $1^\circ \times 1^\circ$ grids with 5 or more cloud-sliced data points in each data product.
- The use of a geometric AMF to convert slant columns to vertical columns assumes the vertical distribution of NO₂ within each 150 layer is relatively constant. Belmonte Rivas et al. (2015) estimated that the difference between the geometric AMF and an AMF that accounts for surface reflectivity, the vertical $NO₂$ profile and atmospheric scattering is $\leq 10\%$ in all layers, except the lowest layer in that work of 770-870 hPa. In this lowest layer, equivalent to the top half of the boundary layer in our work, the difference in AMFs is up to \sim 30%. The largest differences occur over land where NO_x emissions from sources such as urban traffic, industry, soils, and open burning of biomass cause an exponential increase in NO2 with pressure, unlike over the 155 oceans where the NO2 profile is relatively uniform (Schreier et al., 2015; Wang et al., 2019; Kang et al., 2021; Shah et al.,
- 2023). Given the steep vertical gradient in terrestrial boundary layer $NO₂$, we instead derive $NO₂$ mixing ratios in the lowest layer over terrestrial regions as the difference between TROPOMI seasonal mean cloud-free NO₂ tropospheric columns and free tropospheric columns obtained by integrating cloud-sliced NO2 over the four layers above the boundary layer (800-180 hPa) where data are available in all four overlying layers.
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Cloud fraction and cloud-top height data are from the improved Fast Retrieval Scheme for Clouds from the Oxygen A-band wide (FRESCO) algorithm called FRESCO-wide (Eskes and Eichmann, 2023). FRESCO-wide minimises the difference

between measured and simulated spectra between 757-758, 760-761 and 765-770 nm and is so-called because the third spectral window is wider than the 765-766 nm window used in the previous FRESCO-S algorithm (Wang et al., 2008; Van Geffen et

- 165 al., 2022). The cloud-top pressure retrieved with FRESCO-wide corresponds to an altitude \sim 1 km lower than the physical cloud-top height, as the cloud-top height retrieval assumes clouds are uniform reflective boundaries (Choi et al., 2014; Loyola et al., 2018). Marais et al. (2021) showed that cloud-sliced NO2 is relatively insensitive to the choice of TROPOMI cloud product. Their use of the TROPOMI Retrieval of Cloud Information using Neural Networks Cloud As Layers (ROCINN-CAL) product yielded upper tropospheric $NO₂$ that was only 4-9 ppty more than that from the FRESCO-S product. The small
- 170 difference results from an extratropical latitude-dependent divergence in cloud-top heights between the two products. The reprocessed TROPOMI NO₂ product (v2.3.1) includes data from two cloud retrieval algorithms, FRESCO-wide and O₂-O₂ cloud (O22CLD). FRESCO-wide is used here, as we find that it yields greater data density than the O22CLD product and differences in NO₂ between the two products for coincident grids are small $(< 10\%$). As of August 2023, ROCINN-CAL had not been reprocessed to obtain a consistent record, so is not used.

175 **2.2 NASA DC-8 aircraft observations used to evaluate cloud-sliced NO2**

We evaluate our cloud-sliced NO₂ against NASA DC-8 campaign data. To mitigate interference from decomposition of NO_x reservoir compounds on measured NO₂ over remote regions, we calculate PSS NO₂ for ATom measurements over remote oceans and for all measurements made in the upper troposphere. The PSS $NO₂$ calculation assumes a dynamic daytime equilibrium between NO and NO2 resulting from the balance between photolysis of NO2 yielding NO and reaction of NO with

180 oxidants regenerating NO2. Silvern et al. (2018) estimated with GEOS-Chem that oxidation of NO in the southeast US upper troposphere was mostly (75%) by O₃ followed by the hydroperoxy radical (HO₂) (15%). The remaining 10% is due to oxidation by the methyl peroxy radical (CH₃O₂) and halogen monoxides. Given dominance of O_3 and HO₂ and availability of measurements of these for almost all campaigns used, we calculate PSS NO₂ as follows:

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NO_2 = NO \times \left(\frac{k_1[O_3] + k_2[HO_2]}{j_{NO_2}}\right)
$$
 (1),

where j_{NO_2} is the NO₂ photolysis frequency (in s⁻¹) and k is the rate constant for oxidation of NO by O₃ (k_1) and by HO₂ (k_2) (in cm³ molecule⁻¹ s⁻¹). The square brackets denote concentrations of O_3 and HO_2 in molecules cm⁻³. NO and NO₂ are in pptv. Values of j_{NQ_2} , NO, [O₃], and [HO₂] are from direct measurements and k_1 and k_2 are calculated using the temperature-190 dependent Arrhenius equations documented in the Jet Propulsion Laboratory (JPL) Chemical Kinetics and Photochemical Data publication number 19 (Burkholder et al., 2020). These for cold upper tropospheric temperatures (~220 K) are $k_1 = 1.2 \times 10^{-10}$ ¹⁴ cm³ molecule⁻¹ s⁻¹ and $k_2 = 1.1 \times 10^{-13}$ cm³ molecule⁻¹ s⁻¹. Only aircraft data obtained between 12:00 and 15:00 LST, 1.5

hours around the TROPOMI overpass time of 13:30 LST, are used, to ensure consistent sampling of the midday atmosphere

and that the PSS assumption is valid. We remove aircraft data influenced by stratospheric air, identified as $O₃/CO > 1.25$ mol

- 195 mol⁻¹. We also only use aircraft NO data to calculate PSS NO₂ if the NO measured is double the NO instrument detection limit of 6 pptv. This ensures measurements used are distinct from background noise in our PSS calculation (Ryerson et al., 2000; Yang et al., 2023).
- NASA DC-8 aircraft campaigns with direct observations of $NO₂$ and observations needed to calculate PSS $NO₂$ include 200 INTEX-A in summer 2004 over the United States (Singh et al., 2006), INTEX-B in spring 2006 over the eastern US, the Gulf of Mexico and the northern Pacific Ocean (Singh et al., 2009), ARCTAS in spring and summer 2008 over the Canadian Arctic (ARCTAS Science Team, 2011), SEAC⁴RS in summer and autumn 2013 over the southeast US (SEAC4RS Science Team, 2014), and ATom once in all seasons from 2016 to 2018 following the same pole-to-pole flight path over the Atlantic and Pacific Oceans (ATom Science Team, 2021). Direct NO₂ measurements are from thermal-dissociation laser induced 205 fluorescence (TD-LIF) (Di Carlo et al., 2013) for INTEX-A and INTEX-B, and from chemiluminescence (Ryerson et al., 2000) for all other campaigns. There are other DC-8 aircraft campaigns, such as the Subsonic Assessment Ozone and NOx Experiment (SONEX) over the North Atlantic and the Deep Convective Cloud and Chemistry (DC-3) over the eastern US. These are not included in our comparison, because SONEX has routine influence of stratospheric air (Fuelberg et al., 2000) and because DC-3 targeted thunderstorms with large concentrations of NO_x from lightning, so is not representative of a 210 standard atmosphere (Singh et al., 1999; Barth et al., 2015; Nault et al., 2016). Measurements of HO2 are not available for $SEAC⁴RS$, so the PSS NO₂ calculation for this campaign uses average upper tropospheric $[HO₂]$ from the other three campaigns. We find that INTEX-A measurements of NO yield median PSS NO₂ values at 450-180 hPa that are anomalously
	- large (150-450 pptv) in comparison to PSS NO₂ from SEAC⁴RS (30-130 pptv), so no upper tropospheric (450-180 hPa) INTEX-A values are used.

215 **2.3 The GEOS-Chem chemical transport model**

We use GEOS-Chem to evaluate contemporary knowledge of tropospheric NO_x by comparison to our cloud-sliced $NO₂$ vertical profiles. For this, we use GEOS-Chem version 13.3.4 (https://doi.org/10.5281/zenodo.5764874; accessed 11 May 2022) to calculate 4-year seasonal mean NO2 covering the same vertical ranges as cloud-sliced NO2. Model years sampled (1 December 2015 to 30 November 2019) are different to those for TROPOMI, due to a lag in availability of emission inventory data. The

220 model is driven with NASA Modern Era Retrospective analysis for Research and Applications, version 2 (MERRA-2) reanalysis meteorology at a horizontal resolution of $2^{\circ} \times 2.5^{\circ}$ over 47 vertical layers (30-35 in the troposphere) extending to 0.01 hPa.

Global emissions of all anthropogenic sources except aircraft are from the Community Emissions Data System (CEDS) version 225 2 for 2015 to 2019 (McDuffie et al., 2020). Aircraft emissions of NO_x are from the Aviation Emissions Inventory Code (AEIC) for 2005 (Stettler et al., 2011). We use offline grid-independent soil NO_x emissions from Weng et al. (2020), the online Global

Fire Emissions Database version 4 with small fires (GFED4s) (van der Werf et al., 2017) inventory for open burning of biomass, and offline grid-independent lightning NO_x emissions prepared by Meng et al. (2021) using the parameterisation detailed in Murray et al. (2012).

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- GEOS-Chem exhibits a known underestimate in tropospheric NO2 over global oceans, as evidenced by past studies (Travis et al., 2020; Guo et al., 2023; Shah et al., 2023). We address this by updating the GEOS-Chem chemical mechanism to include photolysis of particle-phase nitrates (pNO₃) liberating NO_x as $NO₂$ and as the reservoir compound nitrous acid (HONO) followed by its prompt photolysis to form NO (Ye et al., 2017; Kasibhatla et al., 2018; Romer et al., 2018; Andersen et al.,
- 235 2023). Photolysis of pNO_3 is implemented in GEOS-Chem by scaling the photolysis of nitric acid (HNO₃) by an enhancement factor (EF). The EF is 100 for coarse-mode pNO₃ and is scaled down using the relative molar concentrations of pNO₃ and sea salt aerosol as in Shah et al. (2023) for fine-mode $pNO₃$. This increases lower tropospheric (≤ 6 km) NO₂ over the remote ocean by up to 15 pptv, but has a smaller effect $(< 10$ pptv increase) above 6 km where $pNO₃$ is much less abundant (Shah et al., 2023). Photolysis of the NO_x reservoir compound peroxypropionyl nitrate (PPN, C₂H₅C(O)OONO₂) leading to formation
- 240 of $NO₂$ occurs in the atmosphere, but is absent in GEOS-Chem. There are no reported laboratory measurements of $NO₂$ quantum yields from PPN. According to the Harwood et al. (2003) laboratory study, PPN absorption cross sections and quantum yields of the nitrate radical (NO₃) are within 10% of peroxyacetyl nitrate (PAN, CH₃C(O)OONO₂), so we use PAN quantum yields and cross sections from Burkholder et al. (2020) to represent PPN photolysis in GEOS-Chem.
- 245 For consistent comparison of the model to cloud-sliced NO2, GEOS-Chem is sampled around the TROPOMI overpass (12:00- 15:00 LST) following a 3-month spin-up from 1 September to 30 November 2015 for chemical initialisation of the 4-year simulation. Tropospheric NO₂ in GEOS-Chem is identified using MERRA-2 tropopause heights and additional filtering is applied to remove stratospheric intrusions (O₃/CO > 1.25 mol mol⁻¹). All-sky model scenes are sampled. Marais et al. (2021) determined by applying cloud-slicing to synthetic columns of NO2 simulated with GEOS-Chem that the difference between
- 250 NO2 under very cloudy and all-sky conditions is small (< 17%). The TROPOMI cloud-sliced data are gridded to the GEOS-Chem grid for the comparison and only grid cells with at least 10 cloud-sliced data points are compared. We use a threshold of 10 to ensure that meaningful comparisons can be made between GEOS-Chem and cloud-slicing without excluding a large number of cloud-sliced data.

3 Results and Discussion

255 **3.1 Vertical distribution of tropospheric NO2 from cloud-slicing TROPOMI**

Fig. 1 shows the spatial distribution of cloud-sliced NO2 in the free troposphere in June-August (JJA) 2018-2021 and December-February (DJF) 2018-2022 and Fig. 2 shows boundary-layer NO2 (below 800 hPa) for the same seasons and years obtained with cloud-slicing over the ocean and differencing over land (Sect. 2.1). The percent filled global $1^\circ \times 1^\circ$ grids is similar in both seasons, though with expected seasonal shifts in regions covered, due to seasonality in the location of clouds

- 260 associated with convective features such as the Intertropical Convergence Zone (ITCZ) and absence of clouds over regions of persistent subsidence west of southern Africa and South America. Coverage is greatest in the mid-troposphere and least at 320-180 hPa. Percent coverage averaged over JJA and DJF is 63% of grid cells for 600-450 hPa and 68% for 800-600 hPa covering most of the tropics, subtropics, and midlatitudes. Slightly fewer (38%) result at 450-320 hPa, decreasing to 8% at 320-180 hPa. The few grid squares that are filled at this height mostly occur in the tropics, due to the higher tropopause and
- 265 greater abundance of optically thick clouds (Wang et al., 1996). In the boundary layer (Fig. 2), a total of \sim 14% of the grids are filled, \sim 11% for direct cloud-slicing and \sim 3% for differencing. The latter is limited to locations over land with cloud-sliced NO2 in the top upper troposphere layer. Per-layer percent grids filled is similar for March-May and September-November.

270 **Figure 1: Seasonal mean NO2 in the free troposphere obtained by cloud-slicing TROPOMI. Columns are June-August (JJA; left) and December-February (DJF; right) multiyear (2018-2021 for JJA, 2018-2022 for DJF) means at 1°** ´ **1°. Rows from top to bottom are 320-180, 450-320, 600-450, and 800-600 hPa. Inset boxes give the number of filled** $1^\circ \times 1^\circ$ **grids. Boundary layer (below 800 hPa) data are in Fig. 2.**

- 275 Throughout the free troposphere in all seasons (Fig. 1), cloud-sliced NO2 is typically 20-60 pptv. In the upper troposphere, lightning NO_x emissions and photolysis of NO_x reservoir compounds sustains $NO₂$ concentrations of 20-70 pptv over the oceans and > 90 pptv over the continents in JJA at 450-320 hPa. NO2 concentrations exceeding 70 pptv in JJA at 450-320 hPa over North America, China and the Indian subcontinent is due to a combination of lightning and convective uplift of surface anthropogenic pollution (Bertram et al., 2007; Hudman et al., 2007). NO2 persists for longer in the cold, dry upper troposphere
- 280 (Ehhalt et al., 1992; Jaeglé et al., 1998; Grewe et al., 2001) than in the mid-troposphere below, so NO₂ concentrations are 20 pptv more over Europe and North America at 450-320 hPa than at 600-450 hPa. NO2 over the open oceans is similar (25-50 pptv) throughout the free troposphere and is due mostly to lightning and continental outflow (Kawakami et al., 1997; Zien et al., 2014). NO2 in excess of 55 pptv over South America and 80 pptv over Central Africa at 800-600 hPa results from a mix of intense continental lightning and seasonal open burning of biomass (Andreae et al., 2001; Christian et al., 2003; Duncan et
- 285 al., 2003). The burning season in South America starts in July and occurs throughout JJA in southern Africa and throughout DJF in Africa north of the tropics (Van Der Werf et al., 2006; Castellanos et al., 2014; Van Der Velde et al., 2021). NO2 is longer-lived in winter, due to cold conditions and slow photolysis (Dickerson et al., 1982; Kenagy et al., 2018), so over continental Europe large surface sources of anthropogenic NO_x and limited lightning activity especially in comparison to the US contribute to 80 ppty more $NO₂$ in DJF than in JJA at 800-600 hPa.

Figure 2: As in Fig. 1, but for the boundary layer (below 800 hPa). Panels are NO2 from cloud-slicing over the oceans (a) and from the differencing approach over land (b) (see Sect. 2.1 for details). Note colourbar ranges differ in panels (a) and (b), and (b) is on a log scale.

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In the marine boundary layer (Fig. 2 (a)), the typical range in $NO₂$ concentrations is similar to the layers above, except close to coastlines influenced by continental outflow of anthropogenic pollution and local NO_x production from busy harbours. Along the east coast of China, for example, NO₂ concentrations are > 90 ppty compared to 25-35 ppty over the remote ocean east of China. Terrestrial boundary layer NO₂ coverage in Fig. 2 (b) is limited to the tropics in JJA and the tropics and southern

- 300 subtropics in DJF where cloud-sliced NO2 data are available in all 4 overlying layers (Fig. 1). In the terrestrial boundary layer, NO2 concentrations exceed 30 pptv and peak at 600 pptv over eastern Brazil in DJF, Central Africa in both seasons and Southeast Asia and the Indo-Gangetic Plain (IGP) in JJA. The peaks in Brazil and Central Africa are due to biomass burning, whereas the peaks in Southeast Asia and the IGP are associated with large urban and industrial sources (Giglio et al., 2010; Ghude et al., 2013; Lu et al., 2024). Steep latitudinal gradients in NO₂ of > 100 ppty obtained with the differencing approach
- 305 for NO2 covering Amazonia and Central Africa is due to influence of intense seasonal burning of savanna-type vegetation bordering dense tropical forests (Chen et al., 2013; Ossohou et al., 2019; Jin et al., 2021; Van Der Velde et al., 2021).

The seasonal mean cloud-sliced NO2 at 450-180 hPa obtained by Marais et al. (2021) that we compare to our data for the same vertical extent and time period (Sect. 2.1) ranges from $NO₂ > 80$ pptv over terrestrial regions to $<$ 50 pptv over remote oceans. 310 The two datasets are spatially consistent in all seasons, yielding Pearson's correlation coefficients (R) of 0.74 in JJA, 0.70 in SON, 0.64 in DJF and 0.65 in MAM. Marais et al. (2021) NO2 is on average 26% more than we obtain with our updated cloud slicing. This difference, decomposed into variance and background using RMA regression, is 25-37% more variance and 17- 22 pptv less background $NO₂$ in our data across all four seasons. The greater background values in Marais et al. (2021) are from susceptibility of their approach to outliers (Sect. 2.1).

Figure 3: Seasonal mean percentage contribution of NO2 each cloud-sliced layer to the tropospheric column. Columns are June-August (JJA; left) and December-February (DJF; right). Rows from top to bottom are 320-180, 450-320, 600-450, 800-600, and 1100- 800 hPa. Data are multivear means at $1^{\circ} \times 1^{\circ}$ **.**

Fig. 3 shows the relative contribution of individual layers to the tropospheric column that is obtained by summing the column densities of cloud-sliced $NO₂$ in each layer for grid cells with data in all layers. This limits coverage to the tropics and subtropics. As expected, the boundary layer contribution is greatest, typically exceeding 55% in locations influenced by intensive anthropogenic activity and biomass burning (Sahu and Sheel, 2014; Beirle et al., 2019; Keita et al., 2021). The

325 relative contribution from layers above the boundary layer exhibit zonal and meridional variability, but are relatively constant with altitude at ~20% over central Africa and ~10% over south Asia.

We also examine the size of interannual variability (IAV) in tropospheric NO₂, according to our cloud-sliced data. This is shown in Fig. 4 for JJA and DJF for a select year (2021 for JJA, December 2020 to February 2021 for DJF), calculated as the

- 330 absolute difference between cloud-sliced NO2 in these years and the multiyear mean (Fig. 1). Only three of the five layers are shown, as coverage is poor for individual years for the other two layers. IAV data are obtained for <1% of all $1^{\circ} \times 1^{\circ}$ grid cells at 180-320 hPa and just 2% in the boundary layer. IAV in the layers shown in Fig. 4 is typically \sim 10 pptv over the remote ocean and \sim 25 pptv over continental regions (eastern US, Europe, tropics). The greater IAV over the continents is due to influence of anthropogenic, open biomass burning and lightning NO_x emissions. IAV $NO₂$ is about 20-50% of the variability
- 335 the multiyear means in Fig. 1 and 2. Relatively large IAV NO2 over the remote oceans is restricted to the edges of sampled areas in the subtropics that have low data density, due to the proximity to regions of persistent subsidence where retrievals from cloud-slicing are not always successful.

340 **Figure 4: Free tropospheric NO2 interannual variability (IAV). Panels are single-year NO2 IAV obtained as the absolute difference between single-year (left: JJA 2021; right: DJF 2020-2021) and multiyear mean cloud-sliced NO2 for the 3 layers with greatest geographic coverage. Only grid squares with at least 5 cloud-sliced data points in the single-year means are compared.**

3.2 Evaluation of cloud-sliced NO2 with observed and calculated (PSS) NO2

- 345 Fig. 5 shows the regions selected to intercompare cloud-sliced and DC-8 NO2 obtained with direct measurements and PSS NO2 (Section 2.2). Selected regions include the North Atlantic Ocean sampled during ATom, the Canadian Arctic sampled during ARCTAS and ATom, the eastern United States sampled during SEAC⁴RS, INTEX-A and INTEX-B, and the Pacific Ocean sampled during ATom and INTEX-B. These regions were chosen to optimise coincidence of aircraft data in all five layers. In many instances, though, coincidence is over a limited extent of the sampling domain, especially the upper troposphere
- 350 in almost all domains and the Pacific Ocean in all layers. Domains sampled in all seasons due to the ATom campaign include the Canadian Arctic and the Pacific and Atlantic Oceans. The most sampled time period is JJA, the greatest regional coverage is over the eastern US, and the mid-tropospheric layers (800-600 and 600-450 hPa) have the most DC-8 data. According to the DC-8 NO₂ data, hotspots (NO₂ > 200 pptv) occur over the US terrestrial boundary layer where there are large surface NO_x emissions. Much lower concentrations of \leq 25 pptv over the remote ocean are due to absence of large local sources.

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Figure 5: Maps of tropospheric NO2 over the north-western hemisphere in June-August for the five cloud-slicing pressure ranges. 360 **Filled circles are DC-8 NO2 data along DC-8 flight tracks (Sect. 2.2). Background values are cloud-sliced NO2. Polygons show the regions sampled for comparison of aircraft and cloud-sliced NO2 in Fig. 6 and 7. These are the North Atlantic, the Canadian Arctic, the eastern United States and the Pacific.**

Fig. 6 and 7 compare median DC-8 and cloud-sliced NO₂ concentrations in MAM and JJA (Fig. 6) and SON and DJF (Fig. 7) 365 for the polygons in Fig. 5. Cloud-slicing data are for 2018-2021 in JJA and SON, 2018-2022 in DJF, and 2019-2022 in MAM. JJA data are compared to the ARCTAS, SEAC⁴RS, INTEX-A and ATom-1 campaigns, DJF to ATom-2, SON to ATom-3 and SEAC4 RS, and MAM to ATom-4, ARCTAS and INTEX-B across all layers except for INTEX-A which we do not use for comparisons in the upper troposphere (450-180 hPa) (Sect. 2.2). Vertical profiles of DC-8 NO₂ are relatively stable (\sim 25-80) pptv) throughout the troposphere over Pacific and North Atlantic Oceans and increase exponentially to \sim 75-450 pptv in the 370 boundary layer over the southeastern US and the Canadian Arctic. Most cloud-sliced $NO₂$ in the mid-troposphere and in the 320-450 hPa layer in the upper troposphere are ≤ 15 pptv different to DC-8 NO₂ in the extensively sampled southeastern US and < 25 pptv in the other locations for medians obtained with more than 5 data points. Greater variability in each layer (wider

interquartile ranges) in either dataset is typically because there are fewer data points and less extensive coverage (Fig. 4).

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Figure 6: Comparison of seasonal mean vertical profiles of DC-8 and cloud-sliced tropospheric NO2. Symbols are median values for the sampling domains in Fig. 5 for data in MAM (open) and JJA (filled). Symbol shapes for both DC-8 and cloud-slicing datasets differentiate medians obtained with ≤ 5 (triangle) and > 5 (circle) data points. Error bars are interquartile ranges (IQR). NO₂ 380 **concentration scales differ and inset boxes in the top row show boundary-layer NO2 exceeding the** *x***-axis range.**

Large differences between DC-8 and cloud-sliced NO₂ occur in the boundary layer and the top tropospheric layer. In these layers, there are few coincident data points (Fig. 5). Most DC-8 data in these 2 layers are over land influenced by ground-based

sources like intense biomass burning in the boundary layer (Alvarado et al., 2010; Bian et al., 2013) and lightning and 385 convective uplift of surface pollution in the upper troposphere, whereas most cloud-sliced $NO₂$ in these 2 layers are over the ocean (Fig. 5). The cluster of points in the boundary layer over New England in the northeast US in Fig. 5 have similar coverage from both datasets. These are on median 30 pptv (IQR: 20-50 pptv) for DC-8 and 25 pptv (IQR: 20-30 pptv) for cloud-sliced NO2. New England is not included in our comparison in Fig. 6 and 7, as sampling over this location is limited to JJA during INTEX-A.

Figure 7: As in Fig. 6, but for SON (open symbols) and DJF (filled symbols).

3.3 Comparison of cloud-sliced vertical profiles to synthetic GEOS-Chem profiles

Fig. 8 shows the percent difference between multiyear mean GEOS-Chem and cloud-sliced NO2 for June-August and

- 395 December-February obtained after regridding the cloud-sliced NO₂ to the GEOS-Chem $2^{\circ} \times 2.5^{\circ}$ grid. Multiyear means in both datasets are compared to minimise influence of interannual variability quantified in Section 3.1. In general, GEOS-Chem NO2 is 30-80% (10-25 pptv) less than cloud-sliced NO2 in remote locations. Specifically, the Southern Ocean in all layers retrieved, South America throughout the free troposphere, and all grid cells except those over Africa in the upper troposphere. Similar spatial patterns and magnitudes of discrepancies to those plotted in Fig. 8 occur in March-May and September-
- 400 November.

Figure 8: Percent difference between cloud-sliced and GEOS-Chem vertical profiles of tropospheric $NO₂$ **. Maps are at** $2^{\circ} \times 2.5^{\circ}$ **. Blue (red) indicates the model is less (more) than the cloud-sliced NO2. Percent difference is calculated as ((GEOS-Chem minus** 405 **cloud-sliced)/cloud-sliced) for all cloud-sliced** $1^\circ \times 1^\circ$ **grid squares filled in each year sampled.**

Inclusion of nitrate photolysis in GEOS-Chem decreases the model underestimate in NO2 over remote regions from 40-80 pptv to on average ~15 pptv in the mid-troposphere. A relatively large model underestimate of 25-40 pptv over oceans may be due

410 to uncertainties in the enhancement factor used to parameterise nitrate photolysis (Section 2.3) (Shah et al., 2023). PPN photolysis is most effective at increasing $NO₂$ in the 2 layers in the upper troposphere where it is abundant and thermally stable, so photolysis dominates its conversion to NO₂. In JJA, for example, PPN photolysis contributes ~65 pptv NO₂ over the northern midlatitudes and isolated enhancements of 50-60 pptv over southeast Asia and extending from Mozambique to Madagascar. As a result of PPN photolysis, the discrepancy between the model and cloud-sliced upper tropospheric NO₂ is relatively small 415 (10-30 pptv) over the terrestrial northern midlatitudes. The model exceeds the cloud-sliced data by 20-50 pptv over the northern midlatitudes at 600-450 hPa during the summer lightning season north of 35°N. These are the latitudes at which lightning NO_x production rates in GEOS-Chem almost double from 260 moles per flash (mol fl⁻¹) to the south to 500 mol fl⁻¹ to the north (Murray et al., 2012). The effect of this on NO₂ is also evident at 450-320 hPa, though the spatial extent and dataset differences are smaller in this layer. 500 mol ft^{-1} prescribed to northern midlatitude lightning far exceeds observationally constrained 420 global mean estimates of \sim 280 mol fl⁻¹ (Marais et al., 2018) and regional mean estimates of 180 mol fl⁻¹ for the northern

midlatitudes (Bucsela et al., 2019) and 230-360 mol fl-1 for the US and western Atlantic (Allen et al., 2021).

The largest differences between the two datasets occur in the boundary layer along coastlines in North America, Europe and China influenced by anthropogenic pollution. This may in part be due to the different years targeted. COVID lockdowns 425 influenced surface emissions of traffic NO_x in the cloud-sliced data and anthropogenic NO_x emissions are steadily declining over North America, Europe and China as a result of air quality regulation (Zhao et al., 2013; Lloret and Valiela, 2016; Clappier et al., 2021). Both COVID lockdowns and emissions reductions policies would contribute to a model overestimate in NO2. GEOS-Chem also exceeds cloud-sliced NO2 at multiple locations in the 800-600 hPa layer. These include southern Africa in JJA and northern Africa in DJF coincident with the dry burning season of these regions, central Asia in all seasons where there 430 are large sources of anthropogenic pollution. The apparent model overestimate over western US at 600-800 hPa occurs in all seasons and may result from a combination of factors. The TROPOMI sampling period includes the high-fire year (2020) (Albores et al., 2023) and the model does not, affecting the comparison in seasons coincident with the fire season (JJA, SON).

The number of cloud-sliced data points are also relatively few over this region of subsidence. It is difficult to diagnose discrepancies in the tropical terrestrial boundary layer, as anthropogenic emissions inventories are prone to misrepresenting 435 sources unique to the tropics (Duncan et al., 2003; Marais and Wiedinmyer, 2016; Vohra et al., 2022) and there are no suitable independent in-situ measurements to validate the differencing approach we use to derive NO2.

4 Conclusions

Global vertical profiles of tropospheric NO₂ were obtained for five discrete layers (180-320 hPa, 320-450 hPa, 450-600 hPa, 600-800 hPa, and below 800 hPa) by cloud-slicing TROPOMI total columns of NO2 above optically thick clouds. These we 440 assessed against directly measured and calculated (photostationary steady-state) NASA DC-8 aircraft NO2 measurements from 2004 to 2018 . We then applied our cloud-sliced NO₂ to evaluate contemporary understanding of climatological tropospheric NO_x as simulated by GEOS-Chem. We found that coverage from cloud-slicing is greatest in the mid-troposphere (60-70%) where there is an abundance of optically thick clouds and least (8% coverage, mostly in the tropics) in the upper troposphere. Cloud-sliced $NO₂$ ranges from \leq 35 pptv throughout the troposphere over remote marine regions, to 20-60 pptv in the free 445 troposphere over continents, to 160-380 pptv in the boundary layer over source regions in the US, Europe and Asia. Free

tropospheric NO₂ exhibits very little interannual variability, ranging from ~10 pptv over oceans to ~25 pptv over land.

We determined from comparison of cloud-sliced NO₂ to NASA DC-8 aircraft observations that cloud-sliced NO₂ differs from DC-8 NO₂ by just 5-15 pptv when sampling in both datasets is abundant and consistent. It was not feasible to assess cloud-

- 450 sliced NO2 in the boundary layer and in the highest cloud-sliced layer, due to a lack of sufficient coincident data in the tropics. The GEOS-Chem model that represents contemporary understanding of tropospheric NO_x simulates $NO₂$ that is typically 10-40 pptv less than cloud-sliced $NO₂$ in the remote upper troposphere and over the remote oceans. This is a substantial improvement on the > 40 pptv model underestimate before accounting for NO_x recycling in the upper troposphere via PPN photolysis and in the middle and lower troposphere via aerosol nitrate photolysis. Differences are greater over source regions
- 455 influenced by lightning and open burning of biomass and with evolving anthropogenic emissions due to rapid development, policies and events like lockdowns in response to the COVID-19 pandemic. A model high bias of 50 pptv over the northern hemisphere mid-troposphere in June-August points to an issue with the model lightning NO_x production rates that are almost double production rates everywhere else.
- 460 Limited coincident reliable observations to validate cloud-sliced $NO₂$ remains a challenge, but as we demonstrate cloud-sliced NO2 hold value for assessing air quality, chemical transport, and Earth System models to identify differences that warrant further investigation, especially given reliance on these models to understand complex tropospheric chemistry, inform policies, and retrieve trace gas abundances from satellites. Geostationary instruments will further enhance the utility of cloud-sliced NO2 datasets to also investigate daytime variability in vertical profiles of tropospheric NOx.

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Data Availability. The multiyear seasonal mean NO₂ from cloud-slicing TROPOMI and from simulating the GEOS-Chem model are publicly available from the UCL Data Repository (https://doi.org/10.5522/04/25782336).

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Author contributions. Study concept by EAM and RPH. RPH led the writing and analysis, simulated GEOS-Chem and cloudsliced TROPOMI NO2 with supervision from EAM. NW provided the NASA DC-8 python processing code. RGR updated the GEOS-Chem model to include PPN photolysis. VS updated the GEOS-Chem model to include particulate nitrate photolysis. All authors reviewed and edited the manuscript.

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Competing interests. The authors declare that they have no conflict of interest.

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