



Assessment of NO₂ uncertainty impact on aerosol optical depth retrievals at a global scale

Akriti Masoom¹, Stelios Kazadzis¹, Masimo Valeri², Ioannis-Panagiotis Raptis^{3,4}, Gabrielle Brizzi², Kyriakoula Papachristopoulou⁵, Francesca Barnaba⁶, Stefano Casadio², Axel Kreuter^{7,8}, Fabrizio Niro⁹

5 ¹Physical-Meteorological Observatory in Davos, World Radiation Center (PMOD/WRC), Davos, 7260, Switzerland

²Serco Italia S.p.A., Frascati, Rome, 00044, Italy

³Institute for Environmental Research and Sustainable Development, National Observatory of Athens (IERSD/NOA), Athens, 15236, Greece

10 ⁴Laboratory of Climatology and Atmospheric Environment, Sector of Geography and Climatology, Department of Geology and Environment, National and Kapodistrian University of Athens, Athens, 15784, Greece

⁵Institute for Astronomy, Astrophysics, Space Applications and Remote Sensing, National Observatory of Athens (IAASARS/NOA), Athens, 15236, Greece

⁶National Research Council, Institute of Atmospheric Sciences and Climate, CNR-ISAC, Rome, 00133, Italy

⁷Institute for Biomedical Physics, Medical University Innsbruck, Innsbruck, 6020, Austria

15 ⁸LuftBlick OG, Innsbruck, 6020, Austria

⁹ESA-ESRIN, Frascati, Rome, 00044, Italy

Correspondence to: Akriti Masoom (akriti.masoom@pmodwrc.ch)

Abstract. This work aims at investigating the effect of NO₂ absorption on aerosol optical depth (AOD) and Ångström exponent (AE) retrievals of sun photometers by synergistic use of the accurate NO₂ characterization for optical depth estimation from co-located ground-based measurements. The analysis was performed for ~7 years (2017-2023) at a global scale for the AOD and AE retrievals by Aerosol Robotic Network (AERONET) sun photometers which uses OMI (Ozone Monitoring Instrument) climatology for NO₂ representation. The deviations in AOD and AE retrievals by NO₂ absorption is accounted for using high-frequency columnar NO₂ measurements by co-located Pandora spectroradiometer belonging to Pandonia Global Network (PGN). The AERONET retrieved AOD was found to be overestimated in half of the cases while also underestimated in other cases as an impact of the NO₂ deviation from “real” (PGN NO₂) values. Over or underestimations are relatively low. About one-third of these stations showed a mean deviation in NO₂ and AOD (at 380 nm and 440 nm) above 0.5×10^{-4} mol-m⁻² and 0.002, respectively, which can be considered as a systematic contribution to the uncertainties of AOD retrievals that are reported to be in the order of 0.01. However, under extreme NO₂ loading scenarios (i.e., 10% highest deviations), even higher AOD deviations were observed that were at the limit or higher than the reported 0.01 uncertainty of the AOD retrieval. The PGN NO₂ based sensitivity analysis of AOD deviation suggested that for PGN NO₂ varying between 2×10^{-4} and 8×10^{-4} mol-m⁻², the median AOD differences were found to rise above 0.01 (even above 0.02) with the increase in NO₂ threshold (i.e., the lower limit from 2×10^{-4} mol-m⁻² to 8×10^{-4} mol-m⁻²). The AOD-derivative product, AE, was also affected by the NO₂ correction (discrepancies between the AERONET OMI climatological representation of NO₂ values and the real PGN NO₂ measurements) on the spectral AOD. The normalized frequency

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distribution of AE (at 440-870 nm and 380-675 nm wavelength pair) was found to be narrower for broader AOD distribution for some stations and vice versa for other stations and a higher relative error at the shorter wavelength (among the wavelength pairs used for AE estimation) lead to a shift in the peak of the AE distribution towards a higher value. Finally, the AOD and AE trends were calculated based on the original AERONET AOD (based on AERONET OMI climatological
40 NO₂) according to the data availability and it was further signified the importance of having a correct (real) NO₂ representation in AOD retrievals as it would possibly impact the respective trends.

1 Introduction

Earth's radiation budget and climate is impacted by both direct and indirect effects of atmospheric aerosols (IPCC, 2021). The direct effect of aerosols is associated with the absorption and scattering of solar radiation (Hobbs, 1993) while the
45 indirect effect involves the interaction of aerosols with clouds by acting as cloud condensation nuclei and potentially altering cloud properties, precipitation, surface fluxes and the energy budget of the atmosphere (Rosenfeld et al., 2014; Herbert and Stier, 2023). Apart from the impact on climate and radiative forcing, aerosols also have adverse effects on human health leading to respiratory, cardiovascular and neurological diseases, hypertension, diabetes and even cancer (Lelieveld et al., 2015; Molina et al., 2020). Aerosol optical depth (AOD) is the most widely used parameter for the estimation of columnar
50 atmospheric aerosol concentrations at different spectral bandwidths.

Sun photometers are passive remote sensing instruments that are used for measuring AOD which is retrieved using the Lambert-Beer law by taking into account the contribution from Rayleigh scattering by atmospheric molecules and absorption by atmospheric constituents like ozone, nitrogen dioxide, water vapor, etc., other than aerosols. The global aerosol networks such as AERONET (Aerosol Robotic Network, <https://aeronet.gsfc.nasa.gov>), SKYNET (<https://www.skynet-isdc.org/aboutSKYNET.php>, Nakajima, T. et al., 2020), GAWPFR (Global Atmospheric Watch – Precision Filter Radiometers, Kazadzis et al., 2018) network use specific methodology to account for the optical depth contributions from
55 these atmospheric constituents in order to retrieve AOD.

AERONET performs optical depth corrections for Rayleigh scattering at all wavelengths, ozone for spectral range 340-675 nm, NO₂ for spectral range 340-500 nm, water vapor for 1020-1640 nm and carbon dioxide and methane for 1640 nm. The
60 uncertainty in AOD retrieval from AERONET algorithm is estimated to be ~0.01 in visible that reaches up to ~0.02 in the UV region (Eck et al., 1999, Giles et al. 2019). Other factors contributing to the AOD uncertainty in different spectral bands include the optical depth estimation from trace gas (ozone, NO₂) absorption which is sensitive to the estimation of the gas concentrations. Specifically, NO₂ absorption is predominant in lower wavelengths (340-500 nm) and hence NO₂ correction is of significant importance at these wavelengths. This enhances the need to investigate the impact of NO₂ absorption based
65 optical depth on AOD retrievals and the possibility of improvements in the retrieval algorithm by a more accurate NO₂ optical depth estimation using ground based NO₂ measurements.



Emission of nitrogen oxides on a global scale from natural sources are more significant than that generated from anthropogenic activities (Seinfeld and Pandis, 2016). The natural sources of NO_x emissions include wildfires, lightning, oxidation of biogenic ammonia and microbial processes in soils. The NO₂ levels due to NO_x emissions from natural sources are referred to as background and are smaller in magnitude in comparison to the anthropogenic NO_x emissions (Koukouli et al., 2022). The NO_x budget is dominated by fossil fuel combustion, biomass burning emissions and anthropogenic activities. Due to inhomogeneous local emission patterns and photochemical destruction in heavy polluted regions, the NO₂ has high spatiotemporal variations and a shorter lifetime having regional confinement near its source (Richter et al., 2005; Boersma et al., 2008; Tzortziou et al., 2014, 2015; Drosoglou et al., 2017; Fan et al., 2021). The high spatiotemporal variation of tropospheric NO₂ can produce significant bias in the AOD retrievals (Arola and Koskela, 2004; Boersma et al., 2004). Therefore, the regions with high tropospheric NO₂ emissions will have a higher likelihood for deviation from the climatological mean values (Giles et al., 2019). Furthermore, there can also be significant diurnal variation in NO₂ (Boersma et al., 2008). Hence, the climatological mean NO₂ values might not be able to represent the actual NO₂ loading and spatial distribution in the atmosphere. This in turn tends to produce potential errors in the retrieval of AOD in the spectral regions having significant NO₂ absorption. However, a synergistic assistance from the models, satellite observations, or collocated surface-based measuring instruments capable of providing temporal columnar products of NO₂ can help in the reduction in the associated uncertainty and hence the accuracy of the total column NO₂ optical depth estimation can increase (Herman et al., 2009; Tzortziou et al., 2012). To this direction, Pandora Global Network (PGN) (<https://www.pandora-global-network.org>), which is a global network of Pandora spectroradiometers that are used for trace gas measurements and provide the NO₂ concentration, can be useful. These instruments can be used to provide a good estimation of NO₂ concentration in the atmosphere that can help reduce the uncertainty in AOD retrievals.

Here we try to follow up a previous work by Drosoglou et al. (2023) that analyzed the impact of NO₂ absorption using PGN spectroradiometers based high-frequency columnar NO₂ on AOD, AE and SSA retrievals from AERONET and SKYNET for the Rome (Italy) urban area for a time period of 2017-2022. The NO₂ based AOD correction showed a systematic overestimation of AOD and AE with mean AOD bias of ~0.003 and ~0.002 at 380 nm and 440 nm, respectively for AERONET and quite higher (~0.007) bias for SKYNET and average AE bias of ~0.02 and ~0.05 for AERONET and SKYNET, respectively. However, for high columnar NO₂ concentrations (>0.7 DU), the average AOD bias ranged between 0.009–0.012 for AERONET, and ~0.018 for SKYNET. As this study was limited to only one location, a global analysis is needed to better analyze such NO₂ correction-based bias in AOD retrievals.

The work presented in this manuscript deals with updating the work of Drosoglou et al., 2023, that was based in only one station, and a first attempt to analyze global results where AERONET and PGN instruments are collocated. So more specific investigation is performed on the global scale for evaluating the effect of low-to-high NO₂ loads on the AOD retrievals by ground-based remote sensing in several sites across the globe in order to understand the wider impact of uncertainties introduced in the aerosol retrievals by the NO₂ absorption. In particular, we analyze long term dataset (~7 years) collected in 33 globally distributed sites where co-located measurements of both NO₂ from Pandora spectroradiometers part of PGN and



AOD from AERONET sun photometers are available. Following the Introduction, Section 2 deals with the observational data, and methodology for the co-located stations, the retrieval of the aerosol parameters used for the analysis and trend analysis, followed by Sect. 3, which presents the results and discussions; and finally, Sect. 4 summarizes the findings of this study.

105 **2 Data and Methodology**

2.1 Data

2.1.1 Columnar aerosol properties measurements (AOD and AE)

AERONET provides the datasets of aerosol optical, microphysical, and radiative properties through ground-based passive remote sensing using Cimel sun photometers (<https://www.cimel.fr/solutions/ce318-t/>). It has a centralized data processing and distribution system providing the instrument calibration standardization and data acquisition. AERONET direct sun algorithm data products obtained from Version 3 processing algorithm (Giles et al., 2019) is employed in this work including Level 1.5 AOD retrievals at 380 nm, 440 nm, 500 nm, 675 nm and 870 nm, and AE retrievals at 440-870 nm and 380-675 nm. Level 1.5 data products are cloud-screened and quality assured. AERONET data used in this work covers a time period between 2017-2023 during which synchronous data from the co-located Pandora instrument are also available. For the trend analysis in Section 2.2.3, AERONET AOD data between 2013-2023 is considered. The standard AERONET AOD retrievals are based on the NO₂ optical depth estimation from Ozone Monitoring Instrument (OMI/Aura) Level-3 climatological (here on referred to as OMIc) total NO₂ values at a spatial resolution of 0.25° by 0.25° and for time period between 2004-2013.

2.2.2 Vertical column NO₂ measurements

The total NO₂ column product used in this study is obtained from Pandora spectroradiometers which are part of PGN. Pandora spectroradiometers perform direct solar irradiance and scattered radiance measurements with high temporal resolution in the spectral range of 280-530 nm for the retrieval of tropospheric and total column densities, near-surface concentrations and vertical profiles of atmospheric trace gases (e.g., NO₂, O₃, and HCHO) (e.g., Herman et al., 2009; Tzortziou et al., 2012, 2015). The total column NO₂ densities are retrieved from the direct-sun measurements with ~0.6 nm resolution in the spectral range of 280-530 nm using Blick software Cede (2021). Pandora NO₂ vertical column density (VCD) used in this analysis is obtained from Level 2 datasets that provides column amounts, concentrations, profiles, etc., direct-sun retrieval code “nvs3” and Blick processor version 1.8. From this dataset, total column NO₂ VCD with high (0, 10) and medium (1, 11) quality flags are considered.



2.2.3 Satellite observations

130 Daily tropospheric NO₂ columns are retrieved from OMI/Aura level 3, version 1.1 global data products gridded as 0.25° x 0.25° (<https://www.earthdata.nasa.gov>) for the time period of 2017-2023. The retrieved columnar NO₂ is cloud screened and the average of the global NO₂ during 2017-2023 was obtained to get an overview of the regions with high NO₂ based on OMI satellite data global measurements as presented in Section 2.2.1. These datasets are referred to as OMId (OMI daily) throughout the manuscript.

2.2 Methodology

135 2.2.1 Study locations

Taking into account the PGN stations around the globe and having data availability as specified in Section 2.1.2 (version and retrieval code), we selected the co-located AERONET stations with matching latitude and longitude. For multiple co-located AERONET stations, the station having closest match with PGN station latitude and longitude, continuous data flow and/or larger data availability was selected. By applying this criterion, we identified a total of 33 co-located globally distributed stations to be used for the analysis (Table 1, refer to Table A1 for details regarding station names used by AERONET and PGN and instrument number). These include 11 stations in Europe, 14 in North America and South America, 7 in Asia and 1 in the Middle East (Figure 1). Out of these, 1 station is in the Southern hemisphere (COM), 1 is a Polar station (NYA) and 5 are high altitude (>1000 m above sea level) stations. Figure 1 also reports the OMId satellite based (as described in section 2.1.3) long-term mean of daily NO₂ values between 2017-2023 and this shows that the selected stations cover NO₂ daily mean load representative of conditions ranging from clean (e.g., < 0.2x10⁻⁴ mol-m⁻²) to polluted (e.g., > 1x10⁻⁴ mol-m⁻²).

145 The co-located AERONET and PGN stations have the latitudes of all PGN stations within AERONET latitude ± 0.10° and in most of the cases with the exact same latitudes (Table 1). While the longitudes of the PGN stations are within AERONET longitude ± 0.05° except Toronto (0.28°) which is a high latitude station where 0.30° corresponds to about 6 km (Table 1). Corresponding to every measurement of AERONET (time of measurement), the nearest matching PGN measurement (similar time of measurement) was selected and then the PGN data was time interpolated to the AERONET time stamp. Following this process, we obtained specific comparison data points for each station during the comparison period of 2017-2023 based on the co-incident data availability from AERONET and PGN which are provided in Table 1 (last column).



Table 1: Description of the 33 co-located AERONET and PGN stations. The distance of PGN site from AERONET site is mentioned in brackets with sign.

No.	Location, Country	Code	Station coordinates of AERONET (\pm PGN)			Years with coincident data	Comparison data points
			Latitude ($^{\circ}$)	Longitude ($^{\circ}$)	Altitude (m)		
1	Aldine, USA	ALD	29.90 (+0.00)	-95.33 (+0.00)	20 (-12)	2021-2023	14607
2	Athens, Greece	ATH	37.97 (+0.02)	23.72 (+0.05)	130 (+0)	2018-2021	13089
3	Atlanta, USA	ATL	33.78 (+0.00)	-84.40 (+0.00)	294 (+16)	2023	10547
4	Beijing, China	BEI	40.00 (+0.00)	116.38 (+0.00)	59 (+0)	2021-2023	7211
5	Boulder, USA	BOU	40.04 (-0.05)	-105.24 (-0.02)	1622 (+38)	2021-2023	25428
6	Brunswick, USA	BRW	40.46 (+0.00)	-74.43 (+0.00)	20 (-1)	2022-2023	9073
7	Brussels, Belgium	BRU	50.78 (+0.02)	4.35 (+0.01)	120 (-13)	2020-2023	6325
8	Comodoro, Argentina	COM	-45.79 (+0.01)	-67.46 (+0.01)	49 (-3)	2017-2021	12770
9	Dalanzadgad, Mongolia	DLG	43.58 (+0.00)	104.42 (+0.00)	1470 (-4)	2023	10556
10	Davos, Switzerland	DAV	46.81 (-0.01)	9.84 (-0.01)	1589 (+1)	2017-2023	16773
11	Dhaka, Bangladesh	DHK	23.73 (+0.00)	90.40 (+0.00)	34 (+0)	2023	4347
12	Egbert, Canada	EGB	44.23 (+0.00)	-79.78 (+0.00)	264 (-13)	2018-2020	17075
13	Granada, Spain	GRN	37.16 (+0.00)	-3.60 (+0.00)	680 (+0)	2023	24222
14	Hampton, USA	HAM	37.02 (+0.00)	-76.34 (+0.00)	12 (+7)	2022-2023	14424
15	Helsinki, Norway	HEL	60.21 (-0.01)	24.96 (+0.00)	52 (+45)	2017-2023	8472
16	Houston, USA	HOU	29.72 (+0.00)	-95.34 (+0.00)	65 (-46)	2021-2023	17603
17	Innsbruck, Austria	INN	47.26 (+0.00)	11.38 (+0.00)	620 (-4)	2022-2023	8840
18	Izana, Spain	IZA	28.31 (+0.00)	-16.50 (+0.00)	2401 (-41)	2022-2023	49862
19	Julich/Joyce, Germany	JYC	50.91 (+0.00)	6.41 (+0.00)	111 (-17)	2019-2023	9621
20	La Porte, USA	LPT	29.67 (+0.00)	-95.06 (+0.00)	7 (+15)	2021-2022	8434
21	Lindenberg, Germany	LDB	52.21 (+0.08)	14.12 (+0.00)	120 (+7)	2019-2023	13447
22	Manhattan, USA	MNH	40.82 (-0.01)	-73.95 (+0.00)	100 (-66)	2018-2023	29230
23	Mexico City, Mexico	MXC	19.33 (+0.00)	-99.18 (+0.00)	2268 (+12)	2018-2023	26116
24	New Haven, USA	NHV	41.30 (+0.00)	-72.90 (+0.00)	2 (+2)	2022-2023	14880
25	Ny-Ålesund, Norway	NYA	78.92 (+0.00)	11.92 (+0.01)	7 (+11)	2020-2023	21575
26	Rome, Italy	ROM	41.90 (+0.00)	12.51 (+0.01)	75 (+0)	2017-2023	63760
27	Sapporo, Japan	SPR	43.07 (+0.00)	141.34 (+0.01)	59 (-13)	2022-2023	8586
28	Seoul, South Korea	SOL	37.46 (+0.10)	126.95 (-0.02)	116 (-30)	2021-2023	24693
29	Tel-Aviv, Israel	TEL	32.11 (+0.00)	34.81 (+0.00)	76 (+0)	2021-2023	50680
30	Toronto, Canada	TOR	43.79 (-0.01)	-79.47 (+0.28)	186 (-49)	2019-2023	17692
31	Tsukuba, Japan	TSU	36.11 (-0.04)	140.10 (+0.02)	25 (+26)	2021-2023	17048
32	Ulsan, South Korea	ULS	35.58 (-0.01)	129.19 (+0.00)	106 (-68)	2021-2023	25745
33	Wallops, USA	WAL	37.93 (-0.09)	-75.47 (-0.01)	37 (-26)	2021	7799

155 * USA: United States of America

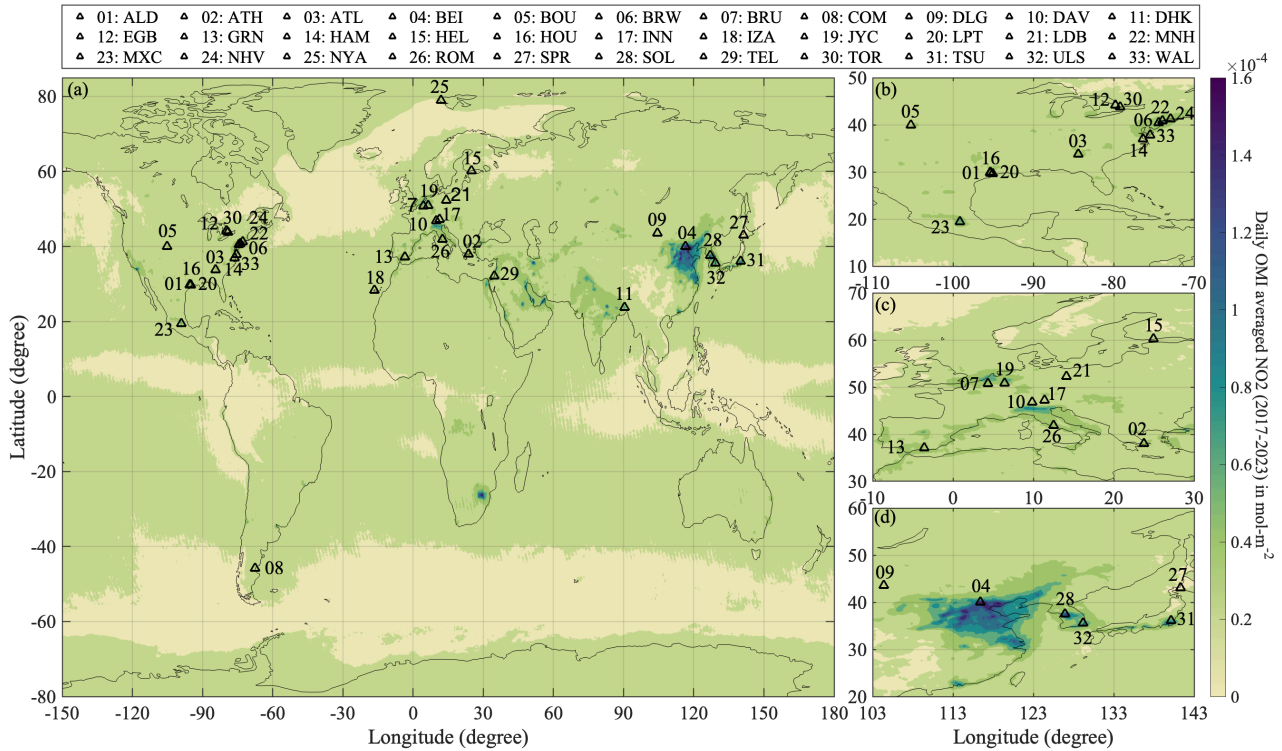


Figure 1: (a) Overview of the co-located AERONET and PGN stations and 7-year (2017-2023) averaged NO₂ (mol·m⁻²) from OMId satellite measurements. Panels (b), (c) and (d) are the focused maps for the clustered locations in North America, Europe and northeast Asia, respectively.

160 2.2.2 NO₂ correction for AOD and AE retrievals

The deviation of the OMIC NO₂ used by AERONET for AOD retrievals from PGN NO₂ VCD (mol·m⁻²) is calculated as

$$\Delta\text{NO}_2 = \text{NO}_{2\text{OMIC}} - \text{NO}_{2\text{PGN}}, \quad (1)$$

where AERONET OMIC NO₂ is converted from Dobson Unit (DU) to SI unit for VCD which is mol·m⁻² (1 DU = 4.4614 × 10⁻⁴ mol·m⁻²) for comparability. AOD is retrieved from direct sun measurements by sun photometers (Cimel sun photometers in case of AERONET) using Lambert–Beer law (Eq. 2) that presents the atmospheric attenuation of radiation as

$$I(\lambda) = I_0(\lambda) * e^{-m\tau} = I_0(\lambda) * e^{-(m_R\tau_R + m_a\tau_a + m_{O_3}\tau_{O_3} + m_{NO_2}\tau_{NO_2})} \quad (2)$$

where $I(\lambda)$ and $I_0(\lambda)$ represent the radiation intensity at surface and top of the atmosphere, respectively at a specific wavelength (λ) and τ is the total optical depth and m being the total optical air mass. Total optical depth is the aggregation of the optical depth contributions from Rayleigh scattering by molecules (τ_R), gaseous absorption by ozone (τ_{O_3}) and NO₂ (τ_{NO_2}) and m_R , m_{O_3} and m_{NO_2} represents their respective optical air masses and m_a is the aerosol optical air mass. The optical air masses are a function of sun elevation. Aerosol optical depth (τ_a) is retrieved from total optical depth (τ) by

subtracting the optical depth contributions from Rayleigh scattering by molecules (τ_R), gaseous absorption by ozone (τ_{O_3}) and NO_2 (τ_{NO_2}). Here, we only discuss about the contribution of NO_2 absorption to AOD and the NO_2 optical depth estimations (Eq. 3) which is calculated as

$$175 \quad \tau_{\text{NO}_2} = \frac{\sigma_{\text{NO}_2}(\lambda)}{1000} * \frac{m_{\text{NO}_2}}{m_a} * \text{NO}_2 \quad (3)$$

where σ_{NO_2} is the NO_2 absorption coefficient at wavelength (λ) and NO_2 VCD is in DU. The NO_2 absorption contribution to the total optical depth is directly proportional to the NO_2 VCD at a specific wavelength and sun elevation. The bias ΔAOD (or $\Delta\tau_a(\lambda)$ as shown in Eq. 5) affecting the AERONET AOD ($\tau_{a,\text{AERONET}}$) retrieval at a specific wavelength produced by the simplified assumption of OMic NO_2 and associated optical depth (which is linear to NO_2 concentration, see Eq. 3) is
 180 evaluated exploiting the ‘real’ value of columnar NO_2 from the co-located PGN instrumentation as shown in Eq. 4 and Eq. 5:

$$\tau_{a,\text{PGN}}(\lambda) = \tau_{a,\text{AERONET}}(\lambda) + \tau_{\text{NO}_2,\text{AERONET}}(\lambda) - \left(\tau_{\text{NO}_2}(\lambda) * \frac{\text{NO}_2\text{PGN}}{\text{NO}_2\text{OMic}} \right) = \tau_{a,\text{AERONET}}(\lambda) - \tau_{\text{NO}_2,\text{AERONET}}(\lambda) \left(\frac{\text{NO}_2\text{PGN}}{\text{NO}_2\text{OMic}} - 1 \right) \quad (4)$$

$$\Delta\tau_a(\lambda) = \tau_{a,\text{AERONET}}(\lambda) - \tau_{a,\text{PGN}}(\lambda) = \tau_{\text{NO}_2,\text{AERONET}}(\lambda) \left(\frac{\text{NO}_2\text{PGN}}{\text{NO}_2\text{OMic}} - 1 \right) = - \frac{\tau_{\text{NO}_2}(\lambda)}{\text{NO}_2\text{OMic}} (\Delta\text{NO}_2) \quad (5)$$

where $\tau_{a,\text{PGN}}$, $\tau_{a,\text{AERONET}}$ and $\tau_{\text{NO}_2,\text{AERONET}}$ represents the PGN NO_2 corrected AOD, original AERONET OMic NO_2 based
 185 AOD and OMic NO_2 based AERONET NO_2 optical depth. Therefore, the sign of the AOD bias depends on the sign of ΔNO_2 i.e., ratio between the OMic and PGN NO_2 . Hence, we define here,

Case 1: OMic NO_2 underestimation, that is $\Delta\text{NO}_2 < 0$ or $\frac{\text{NO}_2\text{PGN}}{\text{NO}_2\text{OMic}} > 1$, leading to a positive AOD bias ($\Delta\text{AOD} > 0$) or overestimation of AOD by AERONET (OMic based AOD) as compared to PGN corrected AOD.

Case 2: OMic NO_2 overestimation, that is $\Delta\text{NO}_2 > 0$ or $\frac{\text{NO}_2\text{PGN}}{\text{NO}_2\text{OMic}} < 1$, leading to a negative AOD bias ($\Delta\text{AOD} < 0$) or
 190 underestimation of AOD by AERONET (OMic based AOD) as compared to PGN corrected AOD.

The spectral variability in AOD is represented by the Ångström exponent (AE) which is obtained from the Ångström power law as:

$$\tau_a(\lambda) = \beta \cdot \lambda^{-\alpha} \quad (6)$$

$$\ln\tau_a(\lambda) = \ln\beta - \alpha \cdot \ln\lambda \quad (7)$$



195 where α and β represents AE and turbidity coefficient, respectively. The negative slope of the least squares regression fit from Equation 7 is used by AERONET to retrieve AE (Eck et al., 1999) with AOD and wavelength for different spectral ranges (here we use 380–675 and 440–870 wavelength pairs for AE estimations) as

$$\alpha_{\lambda_i-\lambda_j} = -\frac{N \sum \ln \tau_i \cdot \ln \lambda_i - \sum \tau_i \cdot \sum \lambda_i}{N \sum (\ln \lambda_i)^2 - (\sum \ln \lambda_i)^2} \quad (8)$$

$\alpha_{\lambda_i-\lambda_j, \text{AERONET}}$ is obtained from AERONET retrieved AE for two wavelength ranges namely 380-675 nm and 440-870 nm. $\alpha_{\lambda_i-\lambda_j, \text{PGN}}$ is calculated from the PGN corrected AOD i.e., $\Delta \tau_{a, \text{PGN}}(\lambda)$ at wavelengths 380 nm and 440 nm (i.e., at λ_i) and from $\Delta \tau_{a, \text{AERONET}}(\lambda)$ at 675 nm and 870 nm (i.e., at λ_j). The difference in the AE is obtained as

$$\Delta \alpha_{\lambda_i-\lambda_j} = \alpha_{\lambda_i-\lambda_j, \text{AERONET}} - \alpha_{\lambda_i-\lambda_j, \text{PGN}} \quad (9)$$

where $\alpha_{\lambda_i-\lambda_j}$ represents the AE in the wavelength range λ_i to λ_j (in our case these wavelength ranges are 380-675 nm and 440-870 nm), $\alpha_{\lambda_i-\lambda_j, \text{AERONET}}$ and $\alpha_{\lambda_i-\lambda_j, \text{PGN}}$ are the AE based on the AERONET AOD and PGN corrected AOD, respectively.

2.2.3 AOD and AE trend estimation

We also evaluate the linear trends in AERONET AOD and AE retrievals for about a decade time span between 2013-2023 to compare them with the mean AOD and AE differences calculated as described in Eq. 5 and Eq. 9. Since, the available PGN data set is for a quite shorter duration for the statistically meaningful calculations of trends, hence we have not considered the trend analysis using PGN corrected AOD and AE.

The linear AOD and AE trends are evaluated using the weighted least squares fitting technique (Weatherhead et al. 1998, Zhang and Reid, 2010; Yoon et al., 2012; Logothetis et al., 2021) as

$$Y_m = \mu + \omega X_m + N_m + S_m, \quad (10)$$

where m represents the index of month ($m = 1, \dots, M$), M is the total number of months, $M/12$ is the total number of years, Y_m represents the monthly average AOD or AE, X_m represents the decimal number of years since the first month of the time series ($m/12$), μ representing a constant linear fit offset at the beginning of the time series, ω represents the magnitude of the respective trend per year, and N_m is the residual. The seasonality is taken into account by subtracting S_m , which is the seasonal term calculated as the long-term monthly mean value, from Y_m . For the purpose of deriving statistically significant daily mean values of the aerosol properties (AOD and AE), a minimum of 10 observations on a daily basis was ascertained. Additionally, in order to have a qualified monthly mean, it was ensured to have the availability of at least 5 days



of measurements on a monthly basis. The data set that did not meet these criteria were not considered in the calculation of AOD and AE trends.

The statistical significance of estimated linear trend (ω) is considered as per the methodology presented by Weatherhead et al. (1998), which has been commonly applied for trend detection in AOD by numerous previous studies (e.g., Ningombam et al., 2019; Zhang et al., 2018; Alfaro-Contreras et al., 2017; Adesina et al., 2016; Pozzer et al., 2015; Kumar et al., 2015, 2018; Li et al., 2014; Babu et al., 2013; Hsu et al., 2012;), by considering N_m that follows a first-order autoregressive process as

$$N_m = \varphi N_{m-1} + \varepsilon_m, \quad (11)$$

where φ is autocorrelation coefficient (lag-1), ε_m represents the white noise and the standard deviation of the trend is calculated as

$$\sigma_\omega \approx \frac{\sigma_N}{n^{3/2}} \sqrt{\frac{1+\varphi}{1-\varphi}}, \quad (12)$$

where σ_N represents the standard deviation of N_m and n is the number of years based on the data availability taking into account the entire period under consideration (i.e., in our case it is a constant value of 11 years). The trends are considered to be significant when the absolute value of ω/σ_ω is above 2.

235 3 Results and Discussion

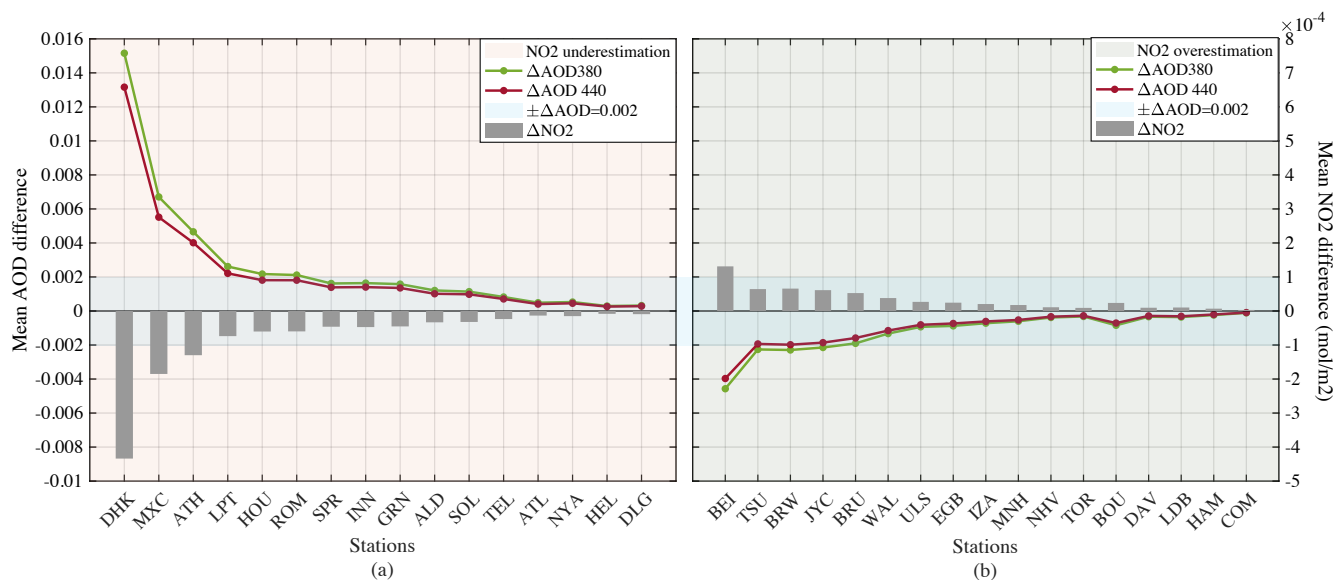
3.1 Differences between NO₂ AERONET OMI climatology and NO₂ PGN measurements and impact on AOD retrievals

As presented in Section 2.2.2, we refer to OMIc NO₂ underestimation (i.e., $\Delta\text{NO}_2 < 0$, PGN/OMIc NO₂ ratio > 1) and hence AOD overestimation ($\Delta\text{AOD} > 0$) as case 1 and OMIc NO₂ overestimation (i.e., $\Delta\text{NO}_2 > 0$, PGN/OMIc NO₂ ratio < 1) leading to AOD underestimation ($\Delta\text{AOD} < 0$) as case 2 which we further discuss here.

Overall, we found 16 (~48% of all the stations) stations in the category of case 1 with mean OMIc NO₂ underestimated as compared to PGN and hence AOD overestimation (Figure 2a). Out of these, 6 stations (DHK, MXC, ATH, LPT, HOU and ROM, ~37%) had mean NO₂ underestimation greater than 0.5×10^{-4} mol-m⁻² and at least 1500 instances with mean $\Delta\text{NO}_2 < -1 \times 10^{-4}$ mol-m⁻² (Appendix Table A2) and, also showed an AOD overestimation equivalent to or above 0.002. For these cases, the corresponding time series of NO₂ values, differences and the normalized frequency distribution of the differences are presented in Figure 3 (panels a-f). The mean PGN and OMIc values in DHK are 5.59×10^{-4} mol-m⁻² and 1.26×10^{-4} mol-m⁻², respectively which has higher “real” (PGN) NO₂ levels reaching even close to 30×10^{-4} mol-m⁻², while OMIc NO₂ remains mostly constant and well within 5×10^{-4} mol-m⁻² (Figure 3a). In ATH, these values are 2.50×10^{-4} mol-m⁻² and 1.20×10^{-4} mol-m⁻², respectively, and in MXC, 3.84×10^{-4} mol-m⁻² and 2.01×10^{-4} mol-m⁻², respectively. These stations also have



250 relatively higher “real” NO₂ values reaching close to 20 x10⁻⁴ mol-m⁻² with OMic NO₂ being mostly constant at ATH and variable at MXC but well within 5 x10⁻⁴ mol-m⁻² for both the stations (Figure 3b and 3c). The corresponding AOD differences at 380 nm are 0.015 (~1.0%), 0.005 (~1.8%) and 0.007 (~1.7%) (Table A2 and Figure A1) for DHK, ATH and MXC, respectively. At 440 nm, these AOD differences are 0.013 (~1%), 0.004 (~1.8%) and 0.005 (~1.7%), for DHK, ATH and MXC, respectively (Figure 2a, Table A2 and Figure A1). The stations LPT and HOU (Figure 1) having an NO₂ difference of 0.71x10⁻⁴ mol-m⁻² and 0.58 x10⁻⁴ mol-m⁻², respectively between OMic and PGN showed a mean difference in AOD as 0.003 and 0.002 (~1.1%) at 380 nm, respectively and 0.002 (~1.1%) at 440 nm. For ROM, ΔNO₂ was found to be -0.60 x10⁻⁴ mol-m⁻² leading to mean AOD overestimation of 0.002 at 380 nm and 440 nm by AERONET OMic as compared to PGN. LPT, HOU and ROM has relatively lesser NO₂ values in time series (reaching close to 10 x10⁻⁴ mol-m⁻² as per Figure 3d, 3e and 3f) as compared to stations like DHK and MXC which are located in high NO₂ zones (as per Figure 1).



260

Figure 2: NO₂ (mol-m⁻²) and AOD differences for all station with NO₂ (a) underestimation and (b) overestimation. The NO₂ differences are calculated as OMic – PGN and the corresponding AOD differences as original AERONET AOD – PGN corrected AOD (as described in Section 2.2.2).

265 The underestimation of NO₂ by AERONET OMic than PGN values at stations like DHK and MXC is possibly due to higher pollution levels which averaged OMic climatological interpretation of NO₂ fails to depict and leads to deviations from the climatological means (Giles et al., 2019). Also, a study by Pavel et al. (2021) on yearly trend analysis of NO₂ for Dhaka showed a statistically significant positive annual slope for the studied period between 2003-2019. Another study over the region of Mexico City using satellite data for time period between 1996-2017 revealed a strong statistically significant positive trend in NO₂ values (Georgoulas et al. 2019).

270 On the other hand, case 2 had 17 (~52% of all the stations) stations with mean NO₂ overestimated by the OMic when compared to PGN leading to AOD underestimation (Figure 2b). Out of these stations, the highest OMic NO₂ overestimation was observed for 4 (~23% of the stations in case 2) stations namely BEI, TSU, BRW and JYC with mean differences above



0.5 $\times 10^{-4}$ mol-m⁻² and at least 1500 instances with the overestimation above 1 $\times 10^{-4}$ mol-m⁻² (Appendix Table A2). These 4 stations also showed the AOD underestimation equal to or above 0.002. The associated NO₂ time series of values, differences and the normalized frequency distribution of the differences can be found in Figure 3 (panels g-j). The average NO₂ values for BEI were found to be 3.06 $\times 10^{-4}$ mol-m⁻² and 4.17 $\times 10^{-4}$ mol-m⁻² from PGN (NO₂ values even reaching close to 20 $\times 10^{-4}$ mol-m⁻², Figure 3g) and OMic, respectively, 1.31 $\times 10^{-4}$ mol-m⁻² and 1.94 $\times 10^{-4}$ mol-m⁻², respectively for TSU, 1.54 $\times 10^{-4}$ mol-m⁻² and 2.16 $\times 10^{-4}$ mol-m⁻², respectively for BRW and 1.75 $\times 10^{-4}$ mol-m⁻² and 2.36 $\times 10^{-4}$ mol-m⁻², respectively for JYC. These differences led to a mean overestimation of NO₂ from OMic as 1.30 $\times 10^{-4}$ mol-m⁻² for BEI and $\sim 0.62 \times 10^{-4}$ mol-m⁻² for TSU, BRW and JYC which led to an AOD underestimation of ~ 0.005 for BEI and ~ 0.002 for TSU, BRW and JYC.

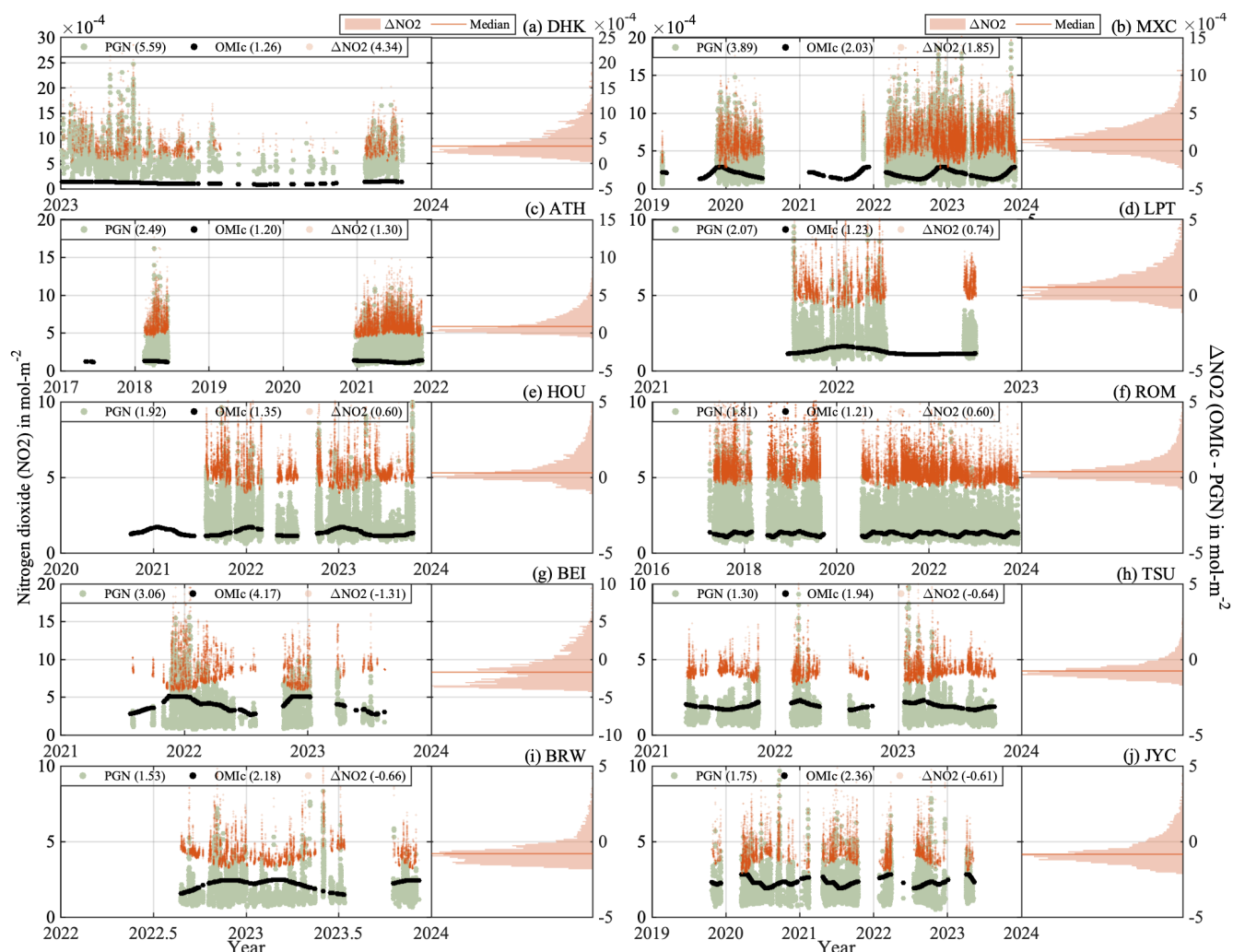


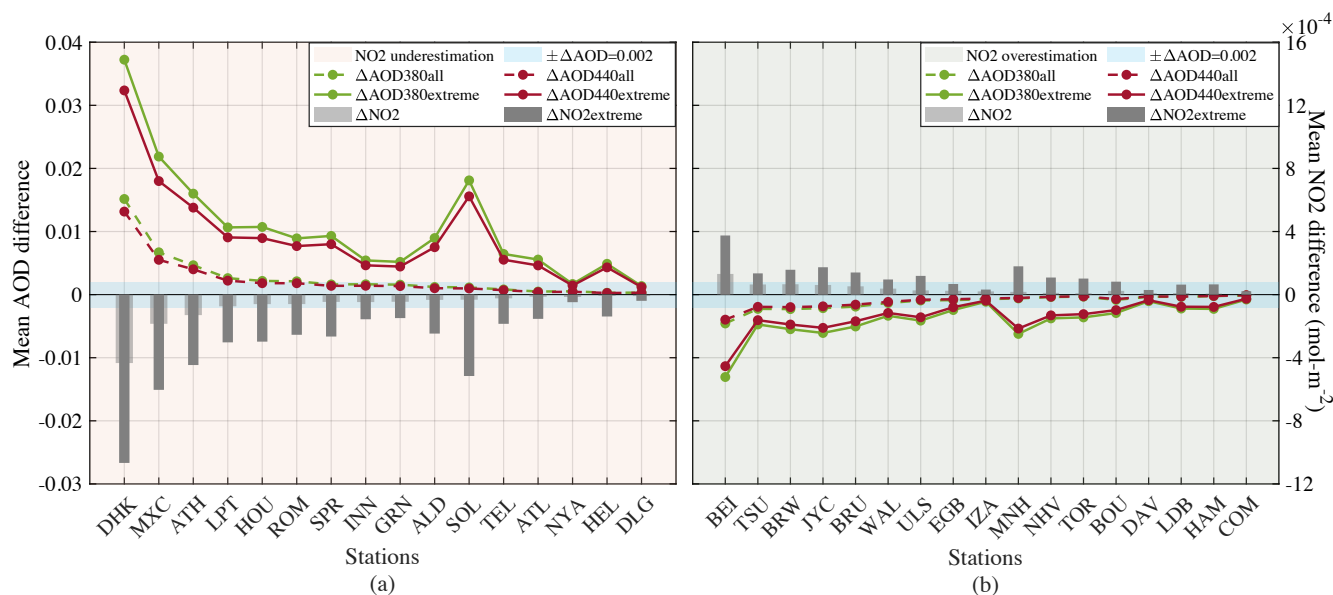
Figure 3: Left panels: Time series of NO₂ (mol-m⁻²) from OMic and PGN, and NO₂ differences (OMic - PGN), Right panels: normalized frequency distribution of the NO₂ differences. The 10 panels refer to stations with mean NO₂ difference above 0.5 $\times 10^{-4}$ mol-m⁻² and mean AOD differences above 0.002. The numbers in the bracket represent the mean values.



Stations like BEI showed an overestimation of NO₂ by AERONET OM1c as compared to PGN possibly due to the reduction in pollution levels as a result of the implementation of environmental protection policies in Eastern China (van der A et al., 2017), that may have led to a significant trend reversal of tropospheric NO₂ during the last decade which OM1c is unable to depict as it considers the average values for time period of 2004–2013.

290 3.2 Assessment of AOD differences in extreme NO₂ load cases

In this section, we present (Table 2) the scenarios with extreme NO₂ situations i.e., 10% highest deviation cases taken into account as percentiles of NO₂ differences with 10% and 90% confidence levels for case 1 (NO₂ underestimation by OM1c) and case 2 (NO₂ overestimation by OM1c), respectively (here on referred to as “Extreme” case). Figure 4 presents a comparison of the NO₂ and AOD differences between the extreme case and whole dataset (referred to as “All”).



295 **Figure 4: Comparison of NO₂ (mol·m⁻²) and AOD differences (OM1c - PGN) in extreme cases with 10% highest NO₂ (a) underestimation and (b) overestimation by OM1c as compared to all datasets.**

Figure 4a presents the results for case 1, in which the mean differences in extreme case were found to be higher than “All” data case for NO₂ by at least 1×10^{-4} mol·m⁻² and 0.003 for AOD for all stations except NYA and DLG. For the 6 selected stations from case 1 as discussed in Section 3.1, this difference between “Extreme” and “All” cases scenario for NO₂ varied from $\sim 2 \times 10^{-4}$ mol·m⁻² reaching up to even 6×10^{-4} mol·m⁻² (for DHK). The increase in AOD differences for these 6 stations was found to be above 0.007 reaching even up to 0.023 and 0.015 for DHK and MXC, respectively. Another station to notice here is SOL, that showed an increase in the average difference in NO₂, AOD380 and AOD440 from 0.34×10^{-4} mol·m⁻², 0.001 and 0.001 in “All” datasets (Fig. 4a) to ~ 15 times (to 5.18×10^{-4} mol·m⁻²), ~ 18 times (to 0.018) and ~ 16 times (to

300



305 0.016), respectively in “Extreme” scenario. Similarly, ALD showed ~7 times and ~8 times increase in the differences in NO₂ and AOD, respectively in “Extreme” scenario as compared to “All” datasets.

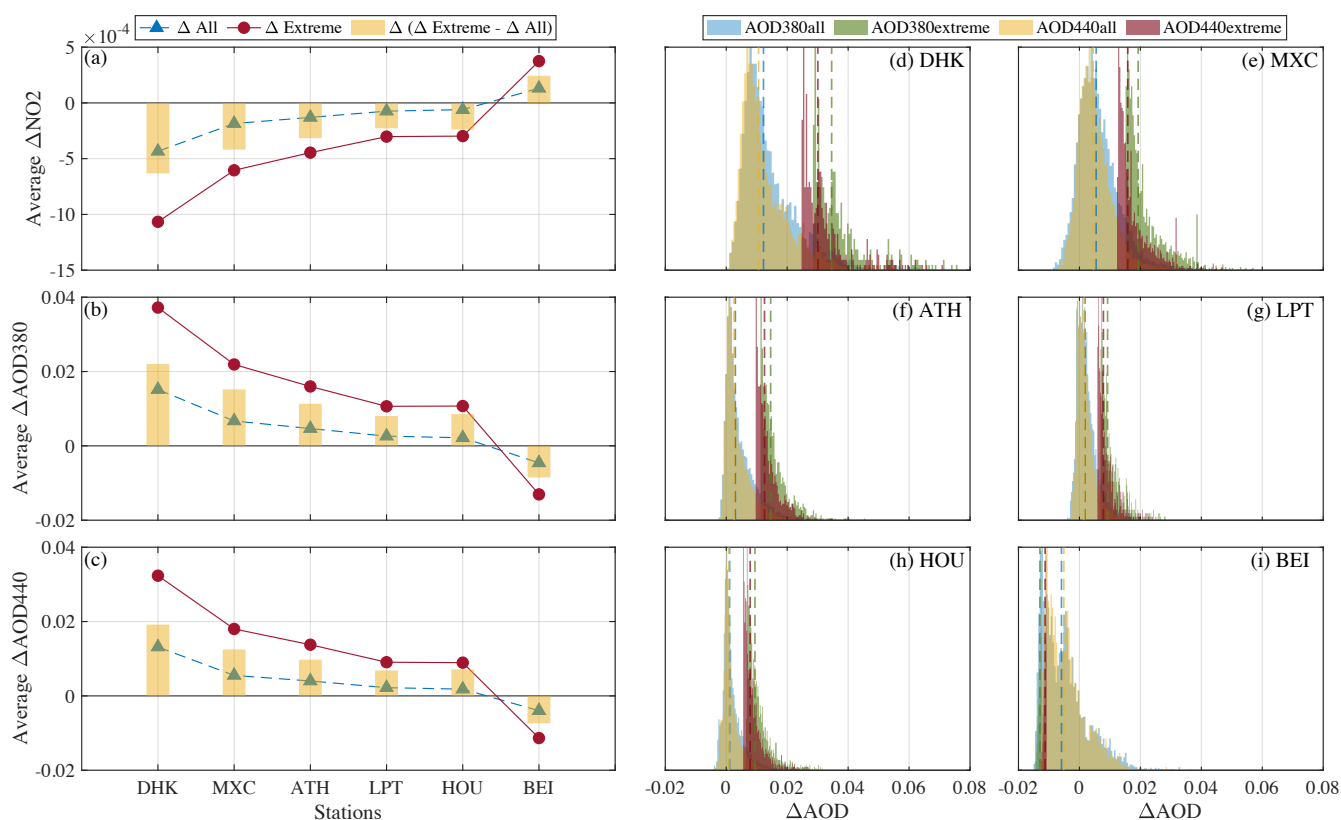
Table 2: Statistics for extreme cases with 10% highest NO₂ deviations (mol-m⁻²) (percentiles (P) at 10% and 90% confidence level for case 1 and case 2, respectively).

Station	ΔNO ₂ All	ΔNO ₂ Extreme	ΔAOD 380 nm		AOD 440 nm		Δ (Average _{Extreme} – Average _{All})			
	x10 ⁻⁴ mol-m ⁻²	x10 ⁻⁴ mol-m ⁻²	Extreme		Extreme					
Case 1: NO ₂ underestimation										
	P (10)	Mean	P (10)	Mean	P (90)	Mean	P (90)	NO ₂	AOD ₃₈₀	AOD ₄₄₀
DHK	-8.23	-10.67	-15.20	0.037	0.053	0.032	0.046	-6.33	0.022	0.019
MXC	-4.27	-6.04	-8.88	0.022	0.032	0.018	0.026	-4.19	0.015	0.012
ATH	-3.19	-4.46	-6.26	0.016	0.022	0.014	0.019	-3.16	0.011	0.010
LPT	-2.00	-3.03	-4.49	0.011	0.016	0.009	0.013	-2.29	0.008	0.007
HOU	-1.89	-2.98	-4.48	0.011	0.016	0.009	0.013	-2.38	0.009	0.007
ROM	-1.55	-2.55	-3.89	0.009	0.014	0.008	0.012	-1.95	0.007	0.006
SPR	-1.52	-2.66	-3.84	0.009	0.013	0.008	0.012	-2.20	0.007	0.007
INN	-1.05	-1.56	-2.31	0.005	0.008	0.005	0.007	-1.09	0.003	0.004
GRN	-1.10	-1.49	-1.97	0.005	0.007	0.004	0.006	-1.04	0.003	0.003
ALD	-1.25	-2.47	-4.69	0.009	0.017	0.008	0.014	-2.14	0.008	0.007
SOL	-3.15	-5.16	-7.62	0.018	0.027	0.016	0.023	-4.84	0.017	0.015
TEL	-1.13	-1.85	-2.84	0.006	0.010	0.006	0.008	-1.61	0.005	0.005
ATL	-0.80	-1.54	-2.60	0.006	0.009	0.005	0.008	-1.41	0.006	0.005
NYA	-0.25	-0.48	-0.96	0.002	0.003	0.001	0.003	-0.33	0.001	0.001
HEL	-0.64	-1.39	-2.37	0.005	0.008	0.004	0.007	-1.31	0.005	0.004
DLG	-0.26	-0.39	-0.57	0.001	0.002	0.001	0.002	-0.30	0.001	0.001
Case 2: NO ₂ overestimation										
	P (90)	Mean	P (90)	Mean	P (10)	Mean	P (10)	NO ₂	AOD ₃₈₀	AOD ₄₄₀
BEI	3.55	3.75	3.96	-0.013	-0.014	-0.011	-0.012	2.44	-0.007	-0.007
TSU	1.22	1.35	1.49	-0.005	-0.005	-0.004	-0.004	0.71	-0.002	-0.002
BRW	1.46	1.58	1.70	-0.005	-0.006	-0.005	-0.005	0.92	-0.002	-0.003
JYC	1.51	1.74	1.93	-0.006	-0.007	-0.005	-0.006	1.13	-0.003	-0.003
BRU	1.23	1.40	1.58	-0.005	-0.006	-0.004	-0.005	0.87	-0.003	-0.002
WAL	0.85	0.96	1.08	-0.003	-0.004	-0.003	-0.003	0.58	-0.002	-0.002
ULS	1.05	1.19	1.33	-0.004	-0.005	-0.004	-0.004	0.92	-0.002	-0.003
EGB	0.56	0.67	0.79	-0.002	-0.003	-0.002	-0.002	0.43	-0.001	-0.001
IZA	0.30	0.32	0.34	-0.001	-0.001	-0.001	-0.001	0.12	-0.000	-0.000
MNH	1.59	1.79	2.00	-0.006	-0.007	-0.005	-0.006	1.61	-0.004	-0.004
NHV	0.92	1.08	1.26	-0.004	-0.004	-0.003	-0.004	0.97	-0.004	-0.003
TOR	0.78	1.02	1.28	-0.004	-0.005	-0.003	-0.004	0.93	-0.003	-0.003
BOU	0.72	0.82	0.92	-0.003	-0.003	-0.002	-0.003	0.58	-0.002	-0.001
DAV	0.24	0.29	0.37	-0.001	-0.001	-0.001	-0.001	0.19	-0.001	-0.001
LDB	0.45	0.63	0.83	-0.002	-0.003	-0.002	-0.003	0.53	-0.002	-0.002
HAM	0.53	0.65	0.76	-0.002	-0.003	-0.002	-0.002	0.58	-0.002	-0.002
COM	0.18	0.22	0.26	-0.001	-0.001	-0.001	-0.001	0.19	-0.001	-0.001

310 For case 2 as presented in Fig. 4b, 8 stations showed the mean deviation between OM_{1c} and PGN NO₂ above 1x10⁻⁴ mol-m⁻² and the deviation of OM_{1c} and PGN NO₂ difference in “Extreme” case from the respective differences in the “All” dataset was found to reach up to ~2 x10⁻⁴ mol-m⁻². These NO₂ deviations lead to an average AOD underestimation of equivalent to or above 0.002 at 380 nm and 440 nm at 14 (out of 17) stations by AERONET. The noticeable station in this case is BEI, JYC and MNH (Fig. 4b) with the deviation of OM_{1c} and PGN NO₂ difference in “Extreme” case from the respective



315 differences in the “All” dataset being above 1×10^{-4} mol-m⁻² leading to higher AOD deviation in “Extreme” case than the “All” dataset by a factor of 0.004 and 0.003 at 380 nm and 440 nm, respectively. It is to be noted that for BEI, the mean AOD underestimation between OMIC and PGN reached to 0.013 and 0.011 at 380 nm and 440 nm, respectively.



320 **Figure 5: ΔNO_2 (mol-m⁻²) (a) and ΔAOD at 380 nm (b) and 440 nm (c) and (d-i) normalized frequency distribution of AOD differences in extreme NO₂ scenario from the whole dataset (referred to as All) for the stations with high variations.**

Figure 5 presents the stations with high variations (AOD deviation of AERONET from PGN equivalent to or above 0.005), the mean NO₂ and AOD differences at these stations as well as the normalized frequency distribution of the AOD at 380 nm and 440 nm. A clear shift of the frequency distribution (Fig. 5d-i) is observed for “Extreme” cases moving away from the “All” dataset case at both wavelengths (380 nm and 440 nm) with larger shift noticeable at DHK and MXC and a shift in
 325 opposite direction in case of BEI which is consistent with the analysis presented in Fig. 4 and Table 2.

Figure 6 presents a sensitivity analysis of AOD differences between AERONET and PGN at 380 nm and 440 nm for all stations with PGN NO₂ varying between 2×10^{-4} and 8×10^{-4} mol-m⁻². The median AOD differences is found to be within ± 0.01 and goes above 0.01 and even above 0.02 with the increase in NO₂ threshold (lower limit) from 2×10^{-4} mol-m⁻² to 8×10^{-4} mol-m⁻². Hence, in case of high NO₂ loadings, the AOD is expected to have higher uncertainties due to inaccurate NO₂
 330 optical depth estimations.

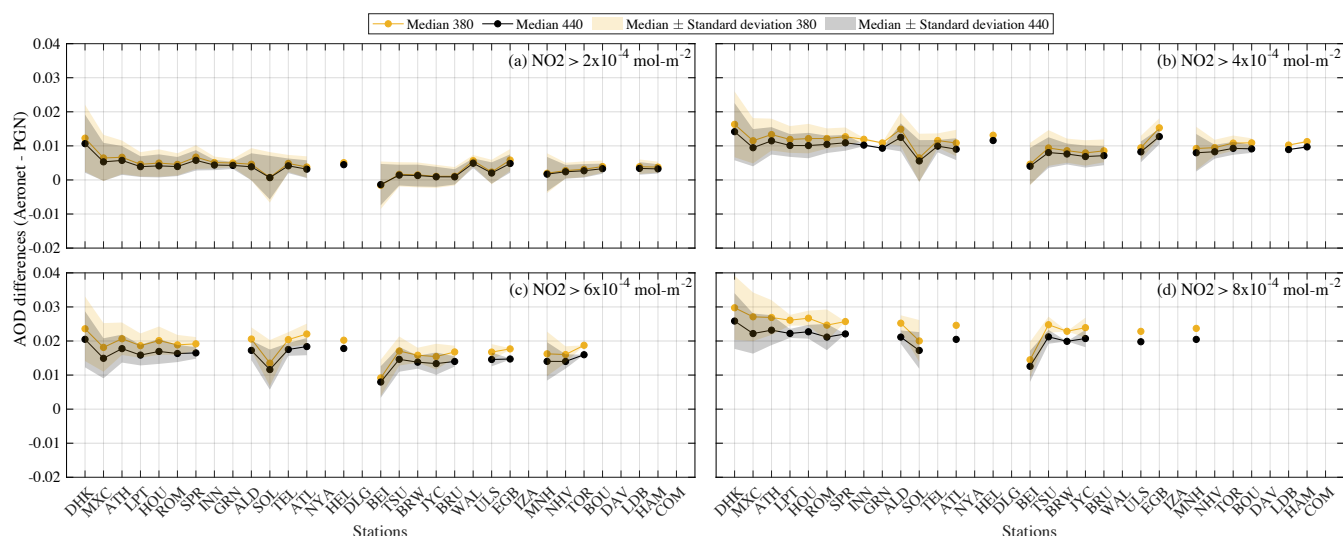
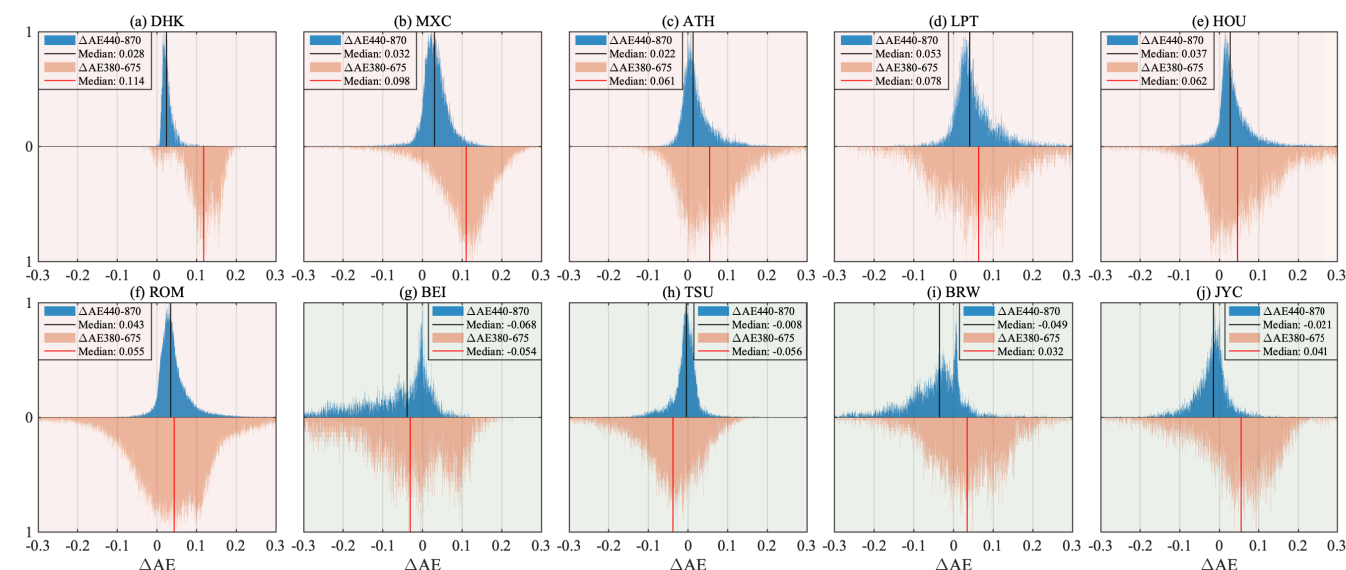


Figure 6: Variation in AOD differences (AERONET OMic based AOD - PGN corrected AOD) at 380 nm and 440 nm for PGN NO_2 varying from (a)-(d) 2×10^{-4} to $8 \times 10^{-4} \text{ mol-m}^{-2}$, respectively for all stations.

3.3 Effect of climatological vs real NO_2 values on AERONET Ångström Exponent

335 Due to a differential impact of the NO_2 correction on the spectral AOD, discrepancies between an assumed climatological NO_2 values (OMic by AERONET) and the real one (PGN based) also impacts the AERONET AOD-based computation of the AE. In this section, we present a discussion regarding the differences in the AERONET AOD based AE and the AE computed from the PGN corrected AOD as is described in Section 2.2.2.



340 **Figure 7:** Normalized frequency distributions of (a-j) the difference in AE at 440-870 nm and 380-675 nm retrieved from the AODs based on AERONET OMic and PGN NO_2 . Shaded background area represents NO_2 underestimation (red) (a-f), and overestimation (green) (g-j) cases.



Figure 7 presents the normalized frequency distribution of these AE differences at the wavelength ranges of 380-675 nm and 440-870 nm. The median of the AE440-870 difference is found to be 0.05 and -0.07 for LPT and BEI, respectively and within ± 0.04 for other stations. The median of the AE380-675 difference is 0.06 for ATH, HOU and ROM, above 0.06 for DHK, MXC and LPT and within 0.05 for the remaining stations. The narrower frequency distribution for stations like DHK can be attributed to the broader AOD distribution (Wagner and Silva, 2008) as shown in Fig. 5d and a broader AE distribution at stations like ATH, LPT, HOU and ROM can be attributed to the narrower AOD distributions at these locations (some examples of AOD distributions are presented in Fig. 5).

In AE retrieval, if the AOD relative errors are equal at both wavelengths, then the AE distribution peak reflects the true value, else there will a shift of the peak of the AE distribution (Wagner & Silva, 2008). In our case, there is no error at higher wavelength (870 nm and 675 nm, as the PGN NO₂ corrections are not made at these wavelengths) and the higher relative error at shorter wavelength (380 nm and 440 nm) leads to a shift in the peak of the AE distribution towards a higher value i.e., the peak of the distribution of AE380-675 is more skewed than that of AE440-870. It is also to be noted that the uncertainty in AE is not very simple to interpret as it is a derivative quantity, and its sensitivity is dependent both on the AOD value as well as any spectral correlations in the AOD uncertainty (Wagner & Silva, 2008; Sayer, 2020).

3.4 Impact of AOD differences on trend analysis

Another aspect of interest relates to the trends in AOD and AE values observed in the last decade, with different magnitude (and even sign i.e., both overestimation and underestimation cases presented in Section 3.1) in different areas of the globe. Hence, in this section, we present the trends based on original AERONET AOD values for a time duration of 2013-2023. In particular, the AOD trends have been calculated based on the AERONET AOD at 380 nm and 440 nm for stations with larger AOD differences ($\Delta AOD > 0.002$) for the time period between 2013-2023, only considering sites with data availability of more than 5 years (complete, i.e., all seasons are homogeneously sampled) over this time span.

Table 3: AERONET AOD trend analysis from 2013-2023 at 380 nm and 440 nm.

Station	No. of Years	AOD 380 nm			AOD 440 nm			AE440-870		
		Trend $\Delta AOD/\text{year}$	Standard error of coefficients	$ \omega/\sigma_\omega $	Trend $\Delta AOD/\text{year}$	Standard error of coefficients	$ \omega/\sigma_\omega $	Trend $\Delta AE/\text{year}$	Standard error of coefficients	$ \omega/\sigma_\omega $
DHK	11	0.011	0.007	1.64	0.009	0.006	1.43	0.01	0.00	3.90
MXC	11	-0.003	0.003	1.11	-0.002	0.002	0.86	-0.00	0.00	0.41
ATH	6	0.000	0.003	0.00	0.000	0.003	0.00	-0.01	0.01	1.81
HOU	11	0.003	0.001	2.15	0.003	0.001	2.40	-0.00	0.01	0.38
ROM	7	-0.001	0.003	0.89	0.001	0.002	0.97	-0.03	0.01	5.63
BEI	11	-0.047	0.005	8.06	-0.036	0.005	6.25	-0.02	0.01	2.70
JYC	11	-0.007	0.002	4.72	-0.006	0.002	4.46	-0.01	0.01	1.84

Table 3 presents the trend analysis using the AERONET AOD and AE. The trends are compared with the mean ΔAOD which was previously presented in Section 3.1. We found two stations with statistically significant negative trends (BEI and



JYC) and one with statistically significant positive trend (HOU) in AOD and negative trends in AE440-870. HOU, having positive AOD trend of 0.003 (Table 3), have mean AOD overestimation of 0.002 at 380 nm and 440 nm (Table A2) which might have impact on the trends when calculated with the corrected AOD values. Furthermore, the other two stations (BEI and JYC) showing a negative trend in AOD showed a mean underestimation of AOD as per the analysis presented in Section 3.1. The trends in AOD were ~ 8 times and ~ 2 times the corresponding mean values for BEI and JYC, respectively. Hence, if the trends can be calculated for these stations with the NO₂ corrected AOD, it may lead to lesser negative trends especially for JYC. The remaining stations (DHK, MXC, ATH and ROM) could not present a statistically significant trends and hence are not discussed here. This analysis signifies the importance of having correct (real) NO₂ values for optical depth calculations that can impact the trend analysis of AOD and AE.

3.5 Pandora NO₂ vertical column density spatial representativeness

In this section, we try to look into the spatial representativeness of the Pandora instruments for the locations as discussed in the previous sections. Figure 8 shows the 7-year averaged OMI satellite values based spatial distribution of NO₂ VCD (also presented in Figure 1) and the statistics are presented in Table 4. The Pandora location (marked in red dots) represents the centre of the circular area (red circles) which are considered according to the OMI satellite overpass (yellow dots). The differences are calculated based on the area averaged NO₂ values from OMI satellite and PAN measurement averages. For stations like DHK and MXC, that have higher NO₂ values, the area averaged differences increase with the increase in the area. While other stations like ATH, LPT, HOU and ROM, showed a comparatively lesser variation in the differences. For BEI, the differences were constants till second circular area around the Pandora site and then increased with the increasing radius and showed maximum difference for the outermost circle.

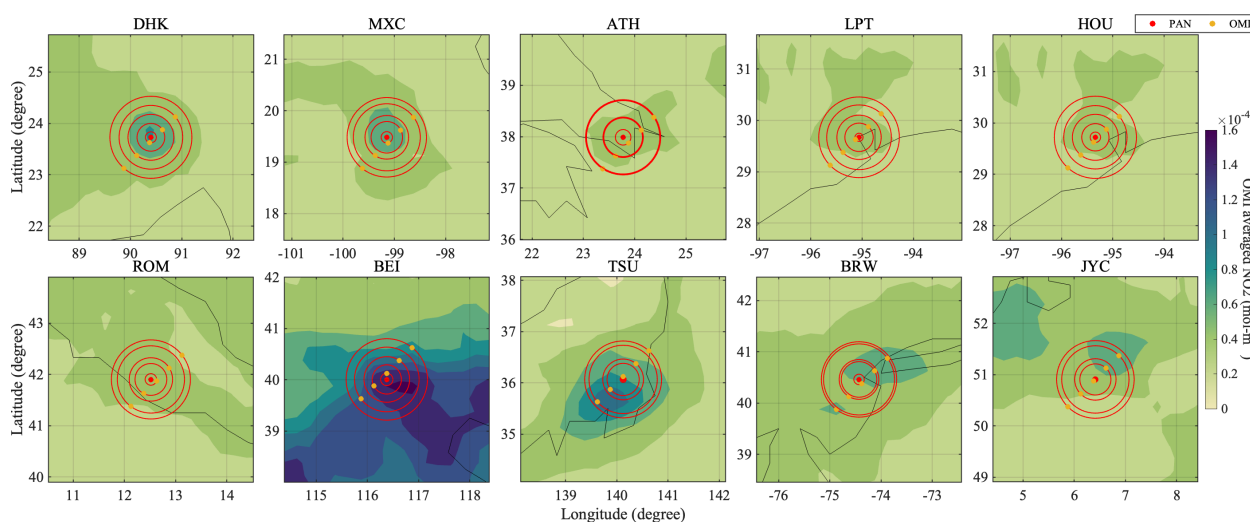


Figure 8: Spatial variation of NO₂ VCD from OMI (7-years averaged value as presented in Figure 1 i.e., during 2017-2023). The red (at the centre) and the yellow dots represents the Pandora location and the satellite overpass, respectively. The red circles centred around the Pandora location are calculated with radius representative of the distance between the Pandora location and satellite overpass.



Table 4: Average NO₂ VCD Pandora – OMI satellite difference in $\times 10^{-4}$ mol-m⁻² circles centred at Pandora site and radius increasing as per the difference between Pandora site and OMI satellite overpass. The circles represent the area around the centre and are numbered according to the increasing distance from the centre. The values in brackets represent the difference of the average NO₂ values of the respective circle from circle 1.

Station	NO ₂ VCD (Pandora – OMI satellite) average difference ($\times 10^{-4}$ mol-m ⁻²)									
	Circle 1		Circle 2		Circle 3		Circle 4		Circle 5	
DHK	4.76	(0.00)	4.86	(0.10)	4.99	(0.23)	5.11	(0.35)	5.22	(0.45)
MXC	3.10	(0.00)	3.19	(0.09)	3.33	(0.22)	3.48	(0.38)	3.54	(0.43)
ATH	2.03	(0.00)	2.04	(0.01)	2.09	(0.06)	2.16	(0.13)	2.19	(0.16)
LPT	1.55	(0.00)	1.61	(0.06)	1.65	(0.11)	1.72	(0.17)	1.76	(0.21)
HOU	1.45	(0.00)	1.44	(-0.01)	1.52	(0.07)	1.58	(0.13)	1.64	(0.18)
ROM	1.31	(0.00)	1.35	(0.04)	1.37	(0.07)	1.48	(0.17)	1.52	(0.22)
BEI	1.58	(0.00)	1.58	(0.00)	1.92	(0.34)	2.05	(0.47)	2.29	(0.71)
TSU	0.50	(0.00)	0.25	(-0.25)	0.51	(0.01)	0.46	(-0.04)	0.65	(0.15)
BRW	0.93	(0.00)	0.74	(-0.19)	0.88	(-0.05)	0.94	(0.01)	0.99	(0.06)
JYC	1.21	(0.00)	1.10	(-0.11)	1.25	(0.04)	1.18	(-0.03)	1.34	(0.13)

395 For sites with homogeneous NO₂ distributions, a pandora instrument can be considered for VCD for larger surrounding area, while for the regions with less homogeneous NO₂ distributions, there can be limited representation of NO₂ in the surrounding area by a pandora instrument (Liu et al., 2024). Moreover, closely located PAN sites like LPT and HOU can be used to include the regional spatial variation in the NO₂. In our analysis, these two closely located stations LPT and HOU (Figure 1) having an NO₂ difference of 0.71×10^{-4} mol-m⁻² and 0.58×10^{-4} mol-m⁻², respectively between OMIc and PGN showed a mean difference in AOD as 0.003 and 0.002 (~1.1%) at 380 nm, respectively and 0.002 (~1.1%) at 440 nm. Another aspect, also shown by Drosoglou et al. (2024) for ATH that analyzed the spatiotemporal variability of NO₂ by synergistically using Pandora and satellite (TROPOMI) observations, could be to use high resolution satellite VCD for NO₂ characterization for real time NO₂ estimations or for the improvement of the climatology used for NO₂ optical depth estimation.

4 Conclusion

405 This work was based on the Drosoglou et al., (2023) findings showing the NO₂ effects on AOD retrievals for Rome, Italy. Here we tried to expand the investigation to all stations with collocated PGN Pandora and AERONET Cimel instruments. We present the analysis of NO₂ deviation between AERONET OMI climatology and PGN dataset focused on the assessment of the impact on AOD at 380 nm and 440 nm from 33 global co-located AERONET and PGN stations. About half of these stations showed an underestimation of NO₂ values by AERONET OMI climatology as compared to the real (PGN) NO₂ measurements that could be possibly due to higher pollution levels which averaged AERONET OMI climatological interpretation of NO₂ fails to depict. While the other stations showed an overestimation of NO₂ which could be possibly due to the reduction in pollution levels as an outcome of the implementation of environmental protection policies (in last decade) that may have led to a significant NO₂ trend reversal which AERONET OMI climatology might not be able to depict due to the fact that it considers the average values for time period of 2004-2013.



415 The correction in AERONET AOD based on PGN NO₂ showed deviation from the AERONET OMI climatology based AOD. The analysis was further focused on 10 stations that showed a minimum mean NO₂ and AOD (at 380 nm and 440 nm) deviations of 0.5×10^{-4} mol-m⁻² and 0.002, respectively. Among these, 10 stations (DHK, MXC, ATH, LPT, HOU and ROM) belonged to case 1 of underestimation of NO₂ and overestimation of AOD, while 4 stations (BEI, TSU, BRW and JYC) showed the overestimation of NO₂ leading to AOD underestimation (case 2).

420 Further assessment of AOD deviations in extreme NO₂ loading scenarios (i.e., 10% highest deviation instances taken into account as percentiles of NO₂ differences with 10% and 90% confidence levels for case 1 and case 2) revealed higher AOD deviations in all cases with much more significant increase in the 10 stations mentioned above along with 3 more stations (ALD, SOL and MNH) as compared to their respective all datasets mean AOD deviations. Furthermore, the sensitivity analysis based on the PGN NO₂ variation from 2×10^{-4} to 8×10^{-4} mol-m⁻² revealed that in case of high NO₂ loadings, the AOD
425 is expected to have higher uncertainties due to inaccurate NO₂ optical depth representation by AERONET OMI climatology.

Due to the impact of the NO₂ correction (discrepancies between the AERONET OMI climatological representation of NO₂ values and the real NO₂ measurements values by PGN) on the spectral AOD, the AOD-derivative product, AE, is also impacted. The normalized frequency distribution of AE was found to be narrower for broader AOD distribution for some stations and vice versa for other stations. A higher relative AOD errors at the shorter wavelength led to the shift in the peak
430 of the AE distribution towards a higher value which is why the peak of the distribution of AE₃₈₀₋₆₇₅ was found to be more skewed than that of AE₄₄₀₋₈₇₀. Also, it is to be noted that the uncertainty in AE is difficult to interpret due to AE being a derivative quantity, and its sensitivity depends both on the AOD value as well as any spectral correlations in the AOD uncertainty.

An AOD and AE trend assessment was made for about a decade for stations with AOD differences above 0.002 and with
435 more than 5 years of data availability based on the original (based on AERONET OMI climatological NO₂) AERONET AOD. On the comparison of the AOD trends with the mean AOD differences (between AOD retrieval based on PGN NO₂ and AERONET OMI climatological NO₂), two stations showed statistically significant negative trends (BEI and JYC) and one showed statistically significant positive trend (HOU) in AOD and negative trends in AE₄₄₀₋₈₇₀. Station having comparable mean AOD overestimation or underestimation with the estimated trends revealed that if the trends can be
440 calculated for these stations with the NO₂ corrected AOD, there can be impacts on the trend values. This analysis signified the importance that a correct (real) NO₂ value could have on the trend analysis of AOD and AE. For future analysis, it would be interesting to see how the NO₂ based AOD correction would impact the AOD and AE trends i.e., how much would the trends deviate when using the corrected AODs.

In general, average AOD related over- or under- estimation due to differences in the actual and climatological NO₂ inputs,
445 are low, with the exception of few stations that satellite based NO₂ climatology fails to capture the local NO₂ variability and its absolute levels. However, in the case of high NO₂ events (days) such deviations are important, as for the top 10% number



of high NO₂ days, for 8 of the stations the impact on AODs is at the limit or higher than the reported 0.01 uncertainty of the AOD retrieval. Taking into account that this uncertainty is a result of various aspects such as: calibration (primarily), post processing and instrument/measurement uncertainty, the NO₂ related contribution can be considered relatively significant.

450 Higher spatial and temporal resolution and updated NO₂ satellite-based climatology or use of collocated Cimel-Pandora retrievals could limit the reported NO₂ related, AOD uncertainties, especially in city areas where NO₂ can be highly variable.

Moreover, some AOD measuring networks (e.g., SKYNET; Nakajima et al., 2020; GAW-PFR; Kazadzis et al., 2018a) do not take officially into account the NO₂ optical depth in AOD retrievals and in this case the NO₂ correction will be considered as a systematic overestimation of AOD. For the GAW-PFR network, NO₂ absorption-based error in AOD

455 retrievals can be assumed to be negligible as the GAW remote stations has low NO₂ concentrations (the annual mean values of NO₂ optical depth are in general < 0.001; Kazadzis et al., 2018a). However, it might be of some significance for stations located in polluted areas specially in Asia or during extreme events such as wildfires which are becoming more frequent as a consequence of climate change. As a future endeavour, it would also be interesting to look into the impact of NO₂ based corrections on AOD and other aerosol properties retrievals especially in ground-based aerosol remote sensing stations

460 located in high pollution zones such as those of SKYNET, which has established regional sub-network groups in China, Europe, India, Japan, South Korea, Mongolia, and Southeast Asia. Finally, the technological improvements and wide spread of instrumentations such as real-time NO₂ monitoring from the Pandonia global network, high spatial resolution real-time satellite-based observations (such as TROPOMI), and the foreseen high temporal resolution NO₂ products (such as from Sentinel 4) could be directly used for contributing towards the improvement of aerosol properties retrievals specifically in

465 the spectral range (~380 – 440 nm) which are significantly affected by NO₂ absorption.



Appendix

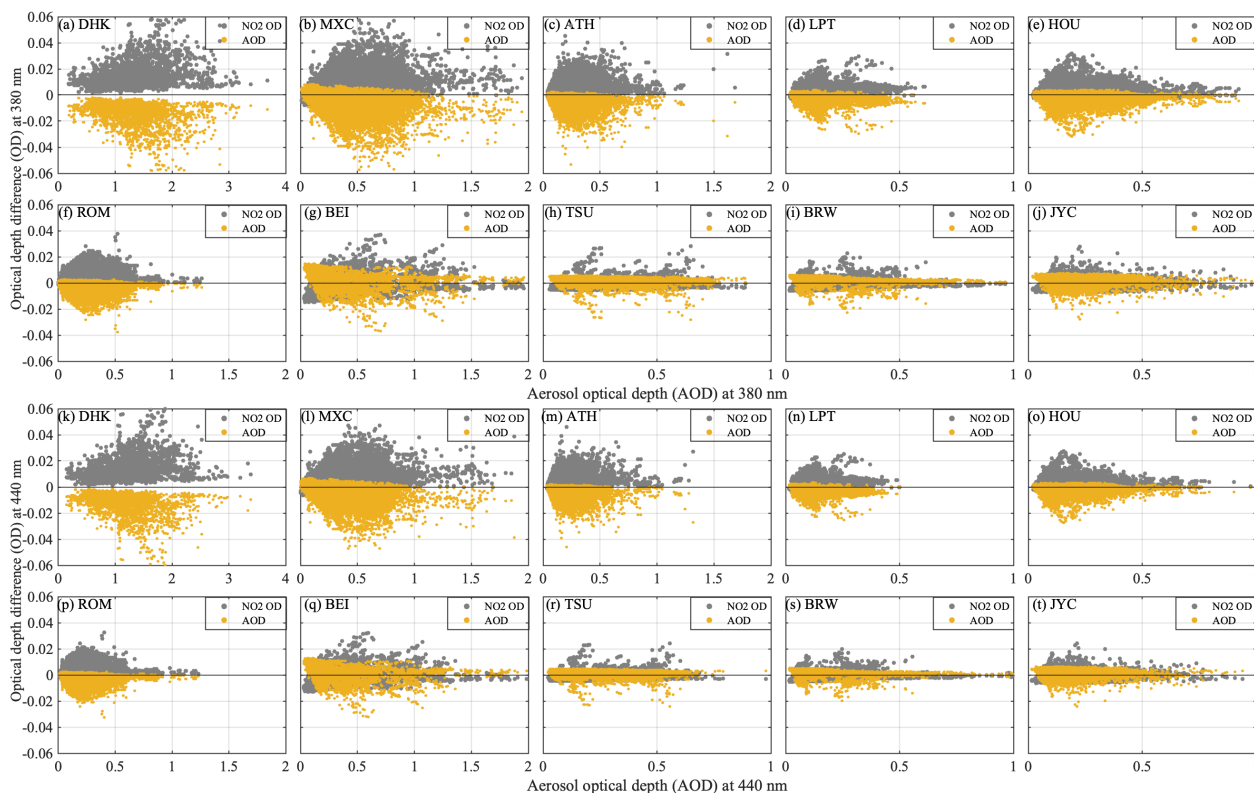
Table A1: AERONET and PGN co-located stations information.

No.	Location, Country	Code	AERONET station name	PGN station name	Pandora instrument no
1	Aldine, USA	ALD	UH_Aldine	AldineTX	61
2	Athens, Greece	ATH	ATHENS-NOA	Athens-NOA	119
3	Atlanta, USA	ATL	Georgia_Tech	AtlantaGA-SouthDeKalb	237
4	Beijing, China	BEI	Beijing_RADI	Beijing-RADI	171
5	Boulder, USA	BOU	NCAR	BoulderCO-NCAR	204
6	Brunswick, USA	BRW	East_Brunswick	NewBrunswickNJ	69
7	Brussels, Belgium	BRU	Brussels	Brussels-Uccle	162
8	Comodoro, Argentina	COM	CEILAP-Comodoro	ComodoroRivadavia	124
9	Dalanzadgad, Mongolia	DLG	Dalanzadgad	Dalanzadgad	217
10	Davos, Switzerland	DAV	Davos	Davos	120
11	Dhaka, Bangladesh	DHK	Dhaka_University	Dhaka	76
12	Egbert, Canada	EGB	Egbert	Egbert	108
13	Granada, Spain	GRN	Granada	Granada	238
14	Hampton, USA	HAM	Hampton_University	HamptonVA-HU	156
15	Helsinki, Norway	HEL	Helsinki	Helsinki	105
16	Houston, USA	HOU	Univ_of_Houston	HoustonTX	25
17	Innsbruck, Austria	INN	Innsbruck_MUI	Innsbruck	106
18	Izana, Spain	IZA	Izana	Izana	209
19	Julich/Joyce, Germany	JYC	FZJ-JOYCE	Juelich	30
20	La Porte, USA	LPT	ARM_LaPorte	LaPorteTX	63
21	Lindenberg, Germany	LDB	MetObs_Lindenberg	Lindenberg	130
22	Manhattan, USA	MNH	CCNY	ManhattanNY-CCNY	135
23	Mexico City, Mexico	MXC	Mexico_City	MexicoCity-UNAM	142
24	New Haven, USA	NHV	New_Haven	NewHavenCT	64
25	Ny-Alesund, Norway	NYA	Ny_Alesund_AWI	NyAlesund	152
26	Rome, Italy	ROM	Rome_La_Sapienza	Rome-SAP	117
27	Sapporo, Japan	SPR	Hokkaido_University	Sapporo	196
28	Seoul, South Korea	SOL	Seoul_SNU	Seoul	54
29	Tel-Aviv, Israel	TEL	Tel-Aviv_University	Tel-Aviv	182
30	Toronto, Canada	TOR	Toronto	Toronto-Scarborough	145
31	Tsukuba, Japan	TSU	TGF_Tsukuba	Tsukuba	193
32	Ulsan, South Korea	ULS	KORUS_UNIST_Ulsan	Ulsan	150
33	Wallops, USA	WAL	Wallops	WallopsIslandVA	40



470 **Table A2: NO₂ (mol-m⁻²), AOD and AE deviations. All differences are as OMic – PGN.**

Station	ΔNO_2 (DU) $\times 10^{-4}$ mol-m ⁻²			ΔAOD 380 nm		ΔAOD 440 nm		ΔNO_2 mol-m ⁻²	ΔAOD	ΔAE				
	Mean	Percentiles		Mean	Percentiles	Mean	Percentiles	cases	cases	Mean	Mean			
Case 1: NO ₂ underestimation														
	50	10		50	90	50	90	< -1x10 ⁻⁴	> 0.01	> 0.005	440-870	440-675		
DHK	-4.34	-3.50	-8.23	0.015	0.012	0.029	0.013	0.011	0.025	4270	2789	4105	0.03	0.05
MXC	-1.85	-1.50	-4.27	0.007	0.005	0.015	0.006	0.005	0.013	16574	6610	13967	0.03	0.04
ATH	-1.30	-0.83	-3.19	0.005	0.003	0.011	0.004	0.003	0.010	5816	1731	4495	0.02	0.04
LPT	-0.74	-0.52	-2.00	0.003	0.002	0.007	0.002	0.002	0.006	2467	357	1538	0.05	0.06
HOU	-0.60	-0.30	-1.89	0.002	0.001	0.007	0.002	0.001	0.006	4044	760	2842	0.04	0.04
ROM	-0.60	-0.38	-1.55	0.002	0.001	0.005	0.002	0.001	0.005	12968	1836	7377	0.04	0.04
SPR	-0.46	-0.15	-1.52	0.002	0.001	0.005	0.001	0.000	0.005	1427	296	943	0.05	0.05
INN	-0.47	-0.35	-1.05	0.002	0.001	0.004	0.001	0.001	0.003	990	22	392	0.04	0.04
GRN	-0.45	-0.31	-1.10	0.002	0.001	0.004	0.001	0.001	0.003	3060	11	1127	0.04	0.03
ALD	-0.33	-0.11	-1.25	0.001	0.000	0.005	0.001	0.000	0.004	1980	400	1266	0.03	0.03
SOL	-0.32	0.15	-3.15	0.001	-0.001	0.011	0.001	-0.000	0.010	7224	2885	5823	0.00	0.00
TEL	-0.24	0.01	-1.13	0.001	0.000	0.004	0.001	0.000	0.003	6046	485	3313	0.01	0.01
ATL	-0.13	-0.03	-0.80	0.000	0.000	0.003	0.000	0.000	0.002	753	88	445	0.02	0.03
NYA	-0.15	-0.12	-0.25	0.001	0.000	0.001	0.000	0.000	0.001	30	0	0	0.02	0.02
HEL	-0.08	0.05	-0.64	0.000	-0.000	0.002	0.000	-0.000	0.002	508	44	304	0.01	-0.01
DLG	-0.09	-0.08	-0.26	0.000	0.000	0.001	0.000	0.000	0.001	6	0	0	0.00	0.00
Case 2: NO ₂ overestimation														
	50	90		50	10	50	10	> 1x10 ⁻⁴	< -0.01	< -0.005	440-870	440-675		
BEI	1.31	1.69	3.55	-0.006	-0.006	-0.012	-0.004	-0.005	-0.011	4660	2023	3929	-0.07	-0.12
TSU	0.64	0.78	1.22	-0.003	-0.003	-0.004	-0.002	-0.002	-0.004	4578	0	358	-0.01	-0.03
BRW	0.66	0.82	1.46	-0.003	-0.003	-0.005	-0.002	-0.002	-0.004	3435	0	1022	-0.05	-0.08
JYC	0.61	0.83	1.51	-0.003	-0.003	-0.005	-0.002	-0.003	-0.005	3591	0	1224	-0.02	-0.04
BRU	0.53	0.63	1.23	-0.002	-0.002	-0.004	-0.002	-0.002	-0.004	1290	0	298	-0.01	-0.03
WAL	0.38	0.34	0.85	-0.001	-0.001	-0.003	-0.001	-0.001	-0.003	295	0	0	-0.01	-0.04
ULS	0.27	0.47	1.05	-0.002	-0.002	-0.004	-0.001	-0.001	-0.003	3157	0	32	-0.01	-0.02
EGB	0.24	0.26	0.56	-0.001	-0.001	-0.002	-0.001	-0.001	-0.002	10	0	0	0.03	0.00
IZA	0.20	0.21	0.30	-0.001	-0.001	-0.001	-0.001	-0.001	-0.001	0	0	0	-0.04	-0.07
MNH	0.18	0.56	1.59	-0.002	-0.002	-0.005	-0.001	-0.002	-0.005	9248	0	4389	-0.01	-0.03
NHV	0.11	0.13	0.92	-0.000	-0.000	-0.003	-0.000	-0.000	-0.003	1002	0	3	-0.02	-0.03
TOR	0.09	0.16	0.78	-0.001	-0.001	-0.003	-0.000	-0.000	-0.002	811	0	88	0.01	-0.01
BOU	0.24	0.27	0.72	-0.001	-0.000	-0.002	-0.001	-0.001	-0.002	12	0	0	-0.03	-0.06
DAV	0.10	0.12	0.24	-0.000	-0.000	-0.001	-0.000	-0.000	-0.001	0	0	0	0.00	-0.01
LDB	0.10	0.07	0.45	-0.000	-0.000	-0.001	-0.000	-0.000	-0.001	0	0	0	0.01	-0.01
HAM	0.07	0.05	0.53	-0.000	-0.000	-0.002	-0.000	-0.000	-0.001	0	0	0	0.01	0.00
COM	0.03	0.05	0.18	-0.000	-0.000	-0.001	-0.000	-0.000	-0.001	0	0	0	0.00	-0.02



475 **Figure A1:** NO₂ optical depth and AOD differences as a function of AOD at (a-j) 380 nm and (k-t) 440 nm for stations with mean NO₂ offset more than $0.5 \times 10^{-4} \text{ mol} \cdot \text{m}^{-2}$ and mean AOD differences offset above 0.002. The numbers in the legend represent the ratio of mean optical depth difference corresponding to NO₂ optical depth (grey) or AOD (yellow) with respect to the mean AOD.

Data availability. The data used in this work are freely available through the AERONET portal at <https://aeronet.gsfc.nasa.gov/> (last access: 26 February 2024), Pandonia global network website at <https://www.pandonia-global-network.org> (last access: 26 February 2024) and NASA Earth Science Data Systems at <https://www.earthdata.nasa.gov> (last access: 26 February 2024).

480 *Author contributions.* AM and SK developed the idea, performed the analysis and prepared the figures. All authors contributed to the discussion of the findings and participated in writing the original manuscript.

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