



1 MAX-DOAS observations of formaldehyde and nitrogen dioxide at three 2 sites in Asia and comparison with the global chemistry transport model 3 CHASER

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12 **Abstract.** Formaldehyde (HCHO) and nitrogen dioxide (NO₂) concentrations and profiles were retrieved
13 from ground-based multi-axis differential optical absorption spectroscopy (MAX-DOAS) observation
14 during January 2017 through December 2018 at three sites in Asia: (1) Phimai in Thailand (15.18°N,
15 102.5°E); (2) Pantnagar (29°N, 78.90°E) in the Indo Gangetic plain (IGP) in India; and (3) Chiba
16 (35.62°N, 140.10°E) in Japan. The NO₂ and HCHO partial columns (<4 km) and profiles simulated using
17 the global chemistry transport model (CTM) and CHASER were compared to those of MAX-DOAS.
18 The vertical sensitivity of the datasets was elucidated using the averaging kernel (AK) information from
19 the MAX-DOAS retrievals. The NO₂ and HCHO concentrations at all three sites showed consistent
20 seasonal variation throughout the investigated period. Biomass burning affected the HCHO and NO₂
21 variation in Phimai during the dry season and in Pantnagar during spring (March-May) and the post-
22 monsoon (September-November) season. High NO₂ concentrations in Phimai during the wet season
23 (June-September) are attributed to soil emissions of nitrogen oxides (NO_x), confirmed from satellite
24 observations and CHASER simulations. Comparison with CHASER shows that the seasonal variations
25 in the HCHO and NO₂ abundances at Phimai and Chiba agree well, with a correlation coefficient (*R*) of
26 0.80. Results agree with the variation, ranging mainly within the one sigma standard deviation of the
27 observations. At Phimai, pyrogenic emissions contribute to the HCHO and NO₂ concentrations up to ~



50 and ~35%, respectively. CHASER showed limited skills in reproducing the NO₂ and HCHO variability at Pantnagar. However, the CHASER simulations in the IGP region showed good agreement with reported results. Sensitivity studies showed that anthropogenic emissions affected the seasonal variation of NO₂ and HCHO concentrations in the IGP region.

1 Introduction

Formaldehyde (HCHO) is the most abundant carbonyl compound in the atmosphere because it is a high-yield product of oxidization of all primary volatile organic carbons (VOCs) emitted from natural and anthropogenic sources by hydroxyl radicals (OH). Oxidization of long-lived VOCs such as methane produces a global HCHO background concentration of 0.2–1.0 ppbv in remote marine environments (Weller et al., 2000; Burkert et al., 2001; Singh et al., 2004; Sinreich et al., 2005). Aside from oxidization of VOCs, the primary emission sources of HCHO are direct emission from biomass burning, industrial processes, fossil fuel combustion (Lee et al., 1997; Hak et al., 2005; Fu et al., 2008;), and vegetation (Seco et al., 2007). However, oxidization of non-methane VOCs emitted from biogenic (e.g. isoprene) or anthropogenic (e.g. butene) sources govern the spatial variability of HCHO on a global scale (Franco et al., 2015). The sinks of HCHO include photolysis at wavelengths below 400 nm, oxidization by OH, and wet deposition, thereby limiting their life to a few hours (Arlander et al., 1995).

Nitrogen dioxide (NO₂), a key and an important atmospheric constituent, (1) participates in the catalytic formation of tropospheric ozone (O₃), (2) acts as a catalyst for stratospheric O₃ destruction (Crutzen, 1970), (3) contributes to the formation of aerosols (Jang and Kamens, 2001), (4) acts as a precursor of acid rain (Seinfeld and Pandis, 1998), and (5) strongly affects radiative forcing (Lelievre et al., 2002; Solomon et al. 1999). Nitrogen oxides (NO_x = NO (nitric oxide) + NO₂) are emitted from natural and anthropogenic sources. Primary NO_x emission sources are biomass burning, fossil fuel combustion, soil emissions, and lightning (Bond et al., 2001; Zhang et al., 2003). Actually, NO_x emissions degrade air quality. For that reason, they are listed as a leading air pollutant with global and national ambient air quality standards (Ma et al., 2013). Both HCHO and NO₂ are important intermediates in the global VOC–HO_x (hydrogen oxides)–NO_x catalytic cycle, which governs the O₃ chemistry in the troposphere (Lee et al., 1997; Houweling et al., 1998; Kanakidou et al., 2005; Hak et al., 2005). Consequently, both trace gases play crucially important roles in tropospheric chemistry.



56 The working principle of multi-axis differential optical absorption spectroscopy (MAX-DOAS), a
57 well-established remote sensing method for measuring trace gases and aerosols, is based on the DOAS
58 technique. Aerosols and trace gases are quantified with analysis of selective narrowband (high frequency)
59 absorption features (Platt 1994, Platt and Stutz 2008). Spectral radiation measurements taken at different
60 elevation angles (ELs) can provide profile information of atmospheric trace gases and aerosols
61 (Hönninger et al., 2004; Wagner et al., 2004; Wittrock et al., 2004; Frieß et al., 2006; Irie et al., 2008a).
62 Many studies have demonstrated the retrieval of aerosol and trace gas concentrations and profiles from
63 MAX-DOAS observations, including the respective concentrations of NO₂ and HCHO (Clemmer et al.,
64 2010; Irie et al., 2011; Hendrick et al., 2014; Franco et al., 2015; Frieß et al., 2016; Wang et al., 2014).

65 The ability of MAX-DOAS to provide information related to surface concentrations, vertical profiles,
66 and column densities makes it complementary to ground-based in situ and satellite observations.
67 Moreover, the MAX-DOAS method uses narrowband absorption of the target compounds, thereby
68 obviating radiometric calibration of the instrument. Because of these advantages, MAX-DOAS systems
69 are deployed for assessment of aerosol and trace gases in regional and global observational networks such
70 as BREDOM (Wittrock et al., 2004), BIRA-IASB (Clemer et al., 2010), and MADRAS (Kanaya et al.,
71 2014). Such datasets are used in, but are not limited to (1) air quality assessment and monitoring, (2)
72 evaluation of chemistry–transport models (CTMS), and (3) validation of satellite retrieval. Several studies
73 have used MAX-DOAS datasets to validate tropospheric columns retrieved from satellite observations,
74 including those of NO₂ and HCHO (; Ma et al., 2013; Irie et al., 2008b; Chan et al., 2020; Ryan et al.,
75 2020). However, limited MAX-DOAS datasets have been used to evaluate global CTMs. Vigouroux et
76 al. (2009) and Franco et al. (2015) respectively used the MAX-DOAS HCHO datasets from Reunion
77 Island and Jungfraujoch stations to evaluate the Intermediate Model of Annual and Global Evolution of
78 Species (IMAGES) and GEOS-Chem model simulations. Kanaya et al. (2014) validated the Model for
79 Interdisciplinary Research on Climate–Earth System Model – Chemistry (MIROC-ESM-CHEM)
80 simulated NO₂ column densities using MAX-DOAS observations in Cape Hedo and Fukue in Japan.

81 For this study, NO₂ and HCHO profiles retrieved from MAX-DOAS observations at the A-SKY
82 (International air quality and sky research remote sensing) (<http://atmos3.cr.chiba-u.jp/a-sky/>) network
83 sites are used to evaluate the global Chemical Atmospheric General Circulation Model for the Study of



84 Atmospheric Environment and Radiative Forcing (CTM CHASER; Sudo et al., 2002). The three A-SKY
85 sites of (1) Phimai in Thailand (15.18°N, 102.56°E), (2) Pantnagar (29°N, 78.90°E) in the Indo Gangetic
86 plain (IGP) in India, and (3) Chiba (35.62°N, 140.10°E) in Japan respectively represent rural, semi-rural,
87 and urban environments. CHASER has been used mostly for global-scale research (Sudo et al., 2007;
88 Sekiya et al., 2014, 2018; Miyazaki et al., 2017). This report is the first of the literature to describe an
89 attempt to evaluate the CHASER-simulated NO₂ and HCHO profiles using MAX-DOAS observations in
90 three atmospheric environments. Moreover, no report of the relevant literature has described the use of
91 MAX-DOAS datasets to evaluate global CTMs in southern and southeastern Asian regions.
92 Consequently, this study was conducted to provide important insights into the model performances and
93 to help reduce model uncertainties related to NO₂ and HCHO simulations in these regions.

94 The manuscript has been compiled in the following manner. The observation sites, MAX-DOAS
95 instrumentation, and retrieval strategies are described in section 2. Section 2 also includes a short
96 description of the CHASER model. Results obtained from the MAX-DOAS observations and the
97 comparison between MAX-DOAS and CHASER datasets are presented respectively in sections 3.1 and
98 3.2. Section 4 presents concluding remarks.

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101 **2 Observations, datasets, and methods**

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103 **2.1 Site Information**

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105 Continuous MAX-DOAS observations at Phimai, Pantnagar, and Chiba started respectively in 2014,
106 2017, and 2012. The MAX-DOAS observations for the two years from January 2017 through December
107 2018 at all three sites are discussed here. Phimai, a rural site, is ~260 km northeast of the Bangkok
108 metropolitan region. Consequently, the Phimai site is unlikely to be affected by vehicular and industrial
109 emissions. The site is, however, affected by two major air streams: the dry, cool northeast monsoon during
110 November – mid-February and the wet, warm southwest monsoon during mid-May – September.



111 Pantnagar, a semi-urban site in India, is located in the Indo-Gangetic Plain region (IGP). The Indian
112 capital of New Delhi is located at ~225 km southwest of the site. The low-altitude plains are in the south
113 and west directions, whereas the Himalayan mountains are located north and east of the site. An important
114 roadway with moderate traffic volume and a small local airport lies within 3 km of the site. Rudrapur (~
115 12 km southwest of Pantnagar) and Haldwani (~. 25 km northeast of Pantnagar) are the two major cities
116 near Pantnagar; a few small-scale industries are established there. Climate classification in Pantnagar is
117 the following: (1) winter (December–February), (2) spring (March–May), (3) summer monsoon (June–
118 August), and (4) autumn (September–November).

119 Chiba, an urban site, is ~. 40 km southeast of the Tokyo metropolitan region. Tokyo Bay, large-scale
120 industries, and residential areas are located within a 50 km radius. The Chiba site is also affected by ship
121 emissions from Tokyo Bay. The Chiba climate is the following: (1) spring (March–May), (2) summer
122 (June–August), (3) autumn (September–November), and winter (December– February). The locations of
123 the three sites are portrayed in Fig. 1.

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Surface HCHO concentrations during June in 2018

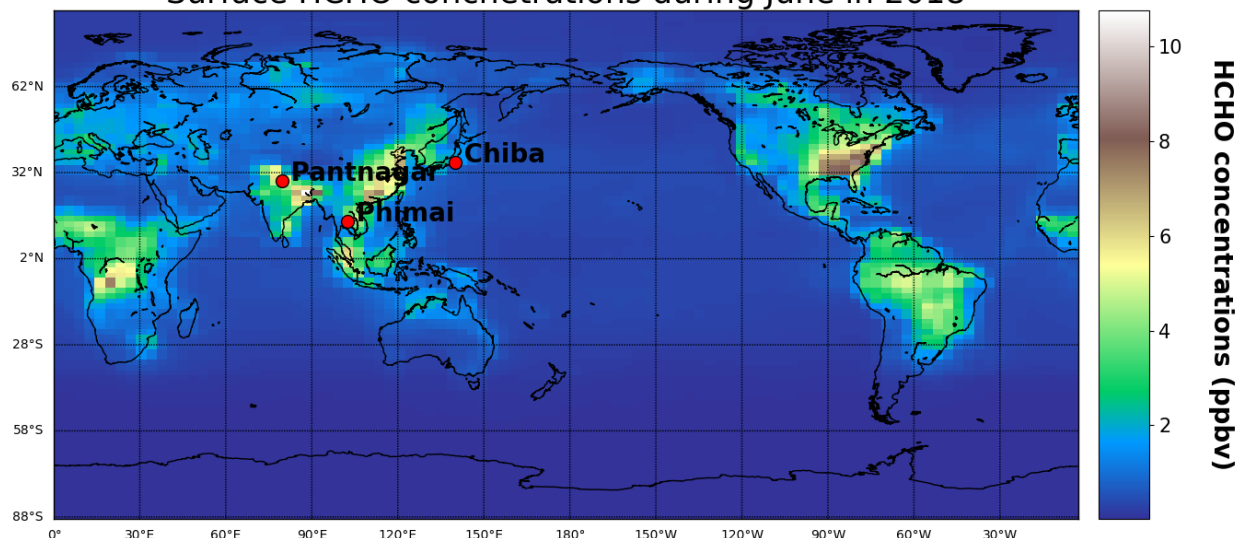


Figure 1: Surface HCHO concentrations during June 2018, simulated using the CHASER model. The red points indicate the locations of the observation sites, which are part of the A-SKY network.

2.2 MAX-DOAS retrieval

The MAX-DOAS systems used for continuous observations at the three sites participated in the Cabauw Intercomparison Campaign of Nitrogen Dioxide measuring Instruments (CINDI) (Roscoe et al., 2010) and CINDI-2 (Kreher et al., 2020) campaign. The instrumentation setup was described by Irie et al. (2008, 2011, 2015). The indoor part of the MAX-DOAS systems consisted of an ultraviolet-visible (UV-VIS)



spectrometer (Maya2000Pro; Ocean Optics Inc.) embedded in a temperature-controlled box. A telescope unit was situated outdoors. The outdoor unit was a single telescope and a 45° inclined movable mirror on a rotary actuator, used to perform reference and off-axis measurements. The high-resolution spectra from 310–515 nm were recorded at six elevation angles (ELs) of 2°, 3°, 4°, 6°, 8°, and 70° at the Chiba and Phimai sites. At the Pantnagar site, measurements were conducted at ELs of 3°, 4°, 5°, 6°, 8°, and 70°. The sequences of the ELs at all the sites were repeated every 15 min. The reference spectra were recorded at EL of 70° instead of 90° to minimize variations in the signals measured at each EL. The off-axis ELs were limited to < 10° to reduce the systematic error in the in-oxygen collision complex (O₄) fitting results (Irie et al., 2015), thereby maintaining high sensitivity in the lowest layer of the retrieved aerosol and trace gas profiles. Daily wavelength calibration using the high-resolution solar spectrum from Kurucz et al. (1984) was performed to account for the spectrometer's probable long-term degradation. The spectral resolution (full width half maximum: FWHM) was about 0.4 nm at 357 and 476 nm. The concentrations and profiles of aerosol and trace gases were retrieved using the Japanese vertical profile retrieval algorithm (JM2 ver. 2) (Irie et al., 2011, 2015). The algorithm works in three steps: (1) DOAS fittings, (2) profile/column retrieval of aerosol, and (3) profile/column retrieval of trace gases. Irie et al. (2008a, 2008b, 2011, 2015) described the retrieval procedures and the error estimates. Here we provide a short overview.

First, the differential slant column density (ΔSCD) of trace gases was retrieved using the DOAS technique (Platt 1994), which uses the nonlinear least-squares spectral fitting method, according to the following equation.

$$\ln I(\lambda) = \ln(I_o(\lambda) - c(\lambda)) - \sum_i^n \sigma_i(\lambda) \Delta\text{SCD}_i - p(\lambda) \quad (1)$$

Therein, $I_o(\lambda)$ represents the reference spectrum measured at time t . Actually, $I_o(\lambda)$ was derived by interpolating two reference spectra (i.e., EL=70°) within 15 min before and after the complete sequential scan of the off-axis ELs at time t . ΔSCD represents the difference between the slant column density along the off-axis and reference spectrum. Second-order and third-order polynomials were fitted to account for



the wavelength-dependent offset $c(\lambda)$ and the effect because of molecular and particle scattering $p(\lambda)$, respectively. Also, $c(\lambda)$ accounts for the influence of stray light. The HCHO Δ SCD and NO_2 Δ SCD are retrieved respectively from the fitting windows of 336–359 nm and 460–490 nm. The O_4 Δ SCD values were retrieved from the fitting windows of 338–370 and 460–490 nm, where significant O_4 absorptions occur. The absorption cross-section data used and the absorbers fitted in the HCHO and NO_2 fitting windows are similar to those of Hoque et al. (2018a, 2018b). Figure 2 presents an example of the fitting results.

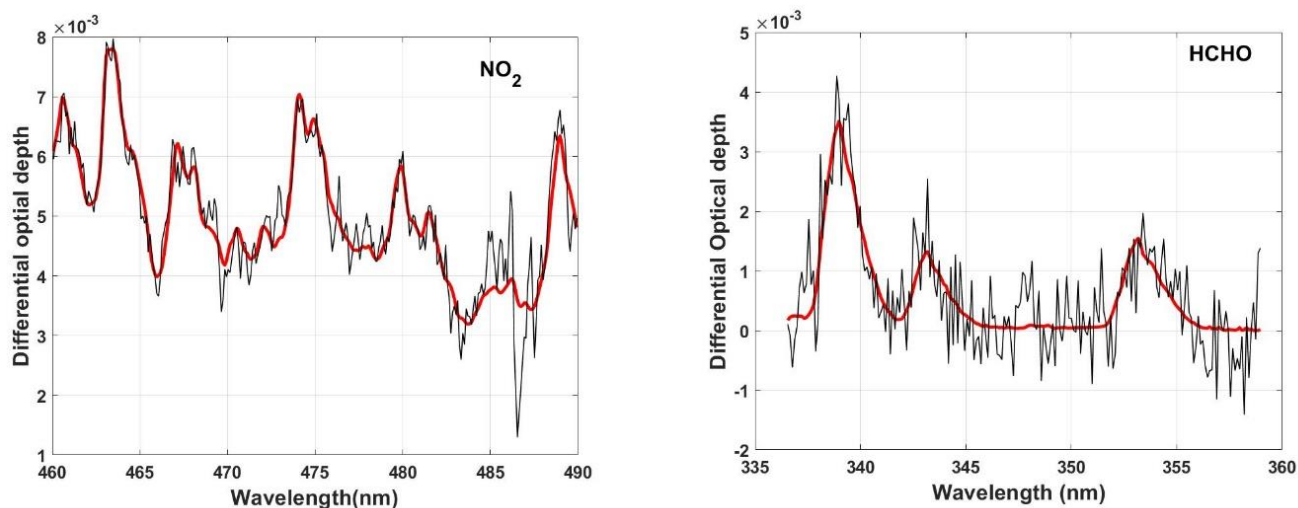


Figure 2: Examples of spectral fitting of NO_2 and HCHO, where the red and black lines show the DOAS fittings and summation of scaled cross-sections and fitting residuals, respectively. The example is shown for the measurements on 10 April 2017, in Phimai at 10:00 LT at an EL of 2° .



In the second step, the aerosol optical depth (AOD) τ and the vertical profiles of the aerosol extinction coefficient (AEC) k were retrieved using the optimal estimation method (Irie et al., 2008a; Rogers, 2000). The measurement vector y (representing the quantities to be fitted) and state vector (representing the retrieved quantities) are defined as

$$y = (O_4 \Delta SCD(\Omega_1) \dots \dots \Delta SCD(\Omega_n))^T \quad (2) \text{ and}$$

$$x = (\tau F_1 F_2 F_3)^T \quad (3),$$

where n stands for the number of measurements within one complete scan of an EL sequence. Also, Ω denotes the viewing geometry, comprising three components: solar zenith angle (SZA), EL, and relative azimuth angle (RAA). The F values determine the profile shape, with values between 0 and 1. The partial AOD for 0–1, 1–2, 2–3, and above 3 km layers were defined respectively as $\text{AOD} \cdot F_1$, $\text{AOD} \cdot (1-F_1) F_2$, and $\text{AOD} \cdot (1-F_1) (1-F_2) F_3$, and $\text{AOD} \cdot (1-F_1) (1-F_2) (1-F_3)$. The AEC profile from layer 3 to 100 km was derived assuming a fixed value at 100 km and exponential AEC profile shape. Similarly, the AEC profiles at 2–3, 1–2, and 0–1 km were derived. Such parameterization provides the advantage that the AEC profile can be retrieved with only the apriori knowledge of the F (profile shape) values and little or no information related to the absolute AEC values in the troposphere. Irie et al. (2008a) demonstrated that the relative variability of the profile shape, in terms of 1-km averages, is smaller than that of the absolute AEC values. However, the vertical resolution and the measurement sensitivity cannot be derived instantly with such parameterization (Irie et al., 2008a; 2009). To overcome such limitations, the retrievals and simulations conducted by other groups for similar geometry (i.e., Frieß et al., 2006) were used. The apriori values used for this study were similar to those reported by Irie et al. (2011): $\text{AOD} = 0.21 \pm 3.0$, $F_1 = 0.60 \pm 0.05$, $F_2 = 0.80 \pm 0.03$, and $F_3 = 0.80 \pm 0.03$.

Then, a lookup table (LUT) of the box air mass factor (A_{box}) vertical profile was constructed using the radiative transfer model JACOSPAR (Irie et al., 2015), which is based on the Monte Carlo Atmospheric Radiative Transfer Simulator (MCARaTS) (Iwabuchi, 2006). Results obtained from JACOSPAR were



validated in the study of Wagner et al. (2007). The optimal aerosol loading and the A_{box} profiles were derived using the A_{box} LUT and the $O_4 \Delta SCD$ at all ELs.

In the third step, the A_{box} profiles, HCHO ΔSCD and $NO_2 \Delta SCD$, and the nonlinear iterative inversion method were used to retrieve the HCHO and NO_2 vertical column densities (VCDs) and profiles. Here the NO_2 retrieval is explained.

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For trace gas retrieval, the measurement vector and state vector were defined as

$$y = (NO_2 \Delta SCD(\Omega_1) \dots \dots NO_2 \Delta SCD(\Omega_n))^T \quad (4) \text{ and}$$

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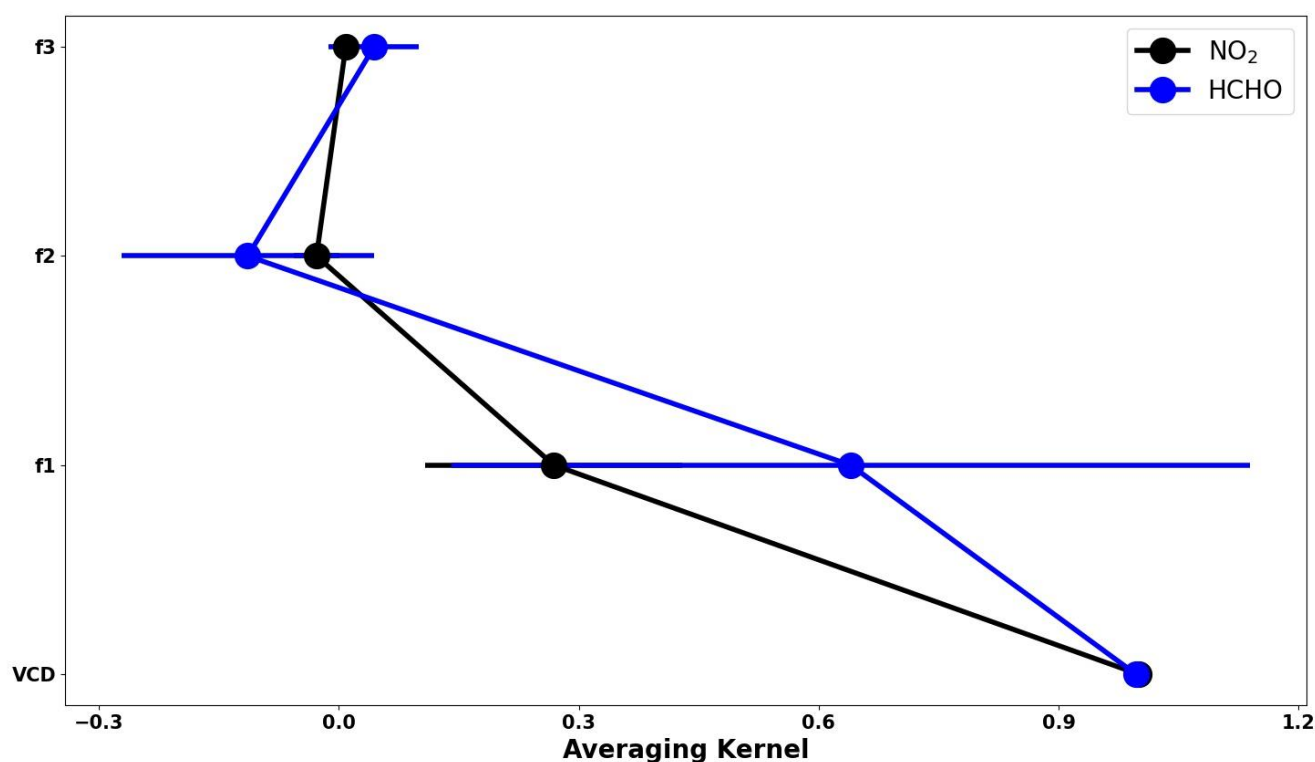
$$x = (VCD f_1 f_2 f_3)^T \quad (5)$$

In those equations, VCD represents the vertical column below 5 km; f values are the profile shape factor. Above the 5 km layer, fixed profiles are assumed. Similarly, to aerosol retrieval, the partial VCD for the 0–1, 1–2, 2–3, and 3–5 km was defined respectively as $VCD \cdot f_1$, $VCD \cdot (1-f_1) f_2$, $VCD \cdot (1-f_1) (1-f_2) f_3$, and $VCD \cdot (1-f_1) (1-f_2) (1-f_3)$. The partial VCD values are converted to the volume mixing ratio (VMR) using the U.S. standard atmosphere temperature, and pressure data scaled to the respective site surface measurements.

The calculated vertical profile was converted to $NO_2 \Delta SCD$ using the A_{box} LUT constructed for aerosol retrieval. However, the trace gas wavelengths differed from the representative wavelengths of A_{box} LUT (357 and 476 nm). The AOD at the trace gas wavelength was estimated, converting the retrieved AOD to the closer aerosol wavelength of 357 or 476 nm, assuming the Angstrom exponent value as 1.00. Then, the A_{box} profiles from the LUT corresponding to the recalculated AOD values were selected. Dependence of the A_{box} profiles on the concentration profiles is expected to be minimal because both HCHO and NO_2 are optically thin absorbers. (Wagner et al., 2007; Irie et al., 2011). For every 15 min (time required for one complete scan of ELs, 20%, the mean ratio of retrieved VCD to maximum ΔSCD) of the maximum trace gas ΔSCD was used as a priori information for VCD retrieval. The a priori error was set to 100% of the maximum trace gas ΔSCD . Figure 3 presents the mean averaging kernel of the HCHO and NO_2 retrievals during the dry season in Phimai. The area calculated from the averaging kernel was close to unity. Therefore, the retrieved VCD was independent of the a priori values.



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237 **Figure 3:** The mean averaging kernel of the NO₂ and HCHO retrievals from the observations in Phimai during
 238 2017

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240 The total error of the retrieval consists of random and systematic errors. The measurement error
 241 covariance matrix constructed from the residuals of the respective trace gas Δ SCDs was used to estimate
 242 the random error. The systematic error was calculated assuming uncertainties as high as 30 and 50% in
 243 the retrieved AOD (or the corresponding A_{box} values). The estimated total error is shown in Table 1. Aside
 244 from the random and systematic error, more sources of error might exist. For instance, the bias in the ELs
 245 can induce uncertainties in the retrieved products. However, Hoque et al. (2018) demonstrated that such
 246 biases had a non-significant effect on the final retrieved products, mostly less than 5%.

247 The cloud screening procedure was similar to that described by Irie et al. (2011) and by Hoque et al.
 248 (2018a, 2018b). During the retrieval steps, retrieved AOD values > 3 were excluded. Optically thick



clouds are primarily responsible for such large optical depth. Furthermore, cloud influence was filtered based on the residuals of O_4 and the trace gas Δ SCDs. The screening criteria were: respective residuals of O_4 , HCHO, and NO_2 of Δ SCD $< 10\%$, $< 50\%$, and $< 20\%$, and the degrees of freedom of the retrieval greater than 1.02.

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Table 1. Estimated Errors (%) for the NO_2 and HCHO concentration in 0-1 km layer, retrieved using theJM2 algorithm

Retrieved Product	Random error	Systematic error	Error related to instrumentation	Total error
NO_2	10	12	5	16
HCHO	16	25	5	30

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2.3 CHASER simulations

CHASER V4.0 (Sudo et al., 2002; Sudo and Akimoto, 2007; Sekiya and Sudo, 2014), coupled online with the MIROC-AGCM atmospheric general circulation model (AGCM) (K-1 model developers, 2004) and the SPRINTARS aerosol transport model (Takemura et al., 2005, 2009), is a global chemistry transport model to study the atmospheric environment and radiative forcing. In addition, several updates, including the introduction of aerosol species (sulfate, nitrate, etc.) and related chemistry, radiation, and cloud processes, have been implemented in the latest version of CHASER.

CHASER can calculate the concentrations of 92 species through 263 chemical reactions (gaseous, aqueous, and heterogeneous chemical reactions) considering the chemical cycle of O_3 – HO_x – NO_x – CH_4 – CO along with oxidation of non-methane volatile organic compounds (NMVOCs) (Miyazaki et al., 2017). CHASER simulates the stratospheric ozone chemistry considering the Chapman mechanisms, catalytic reactions related to halogen oxides (HO_x , NO_x , ClO_x , and BrO_x), and polar stratospheric clouds (PSCs). Resistance-based parameterization (Wesely, 1989), cumulus convection, and large-scale condensation



parameterizations are used to calculate dry and wet depositions. The piecewise parabolic method (Colella and Woodward, 1984) scheme and the flux-form semi-Lagrangian scheme (Lin and Rood, 1996) calculate the advective tracer transport. CHASER simulates tracer transport on a sub-grid scale in the framework of the prognostic Arakawa–Schubert cumulus convection scheme (Emori et al., 2001) and the vertical diffusion scheme (Mellor and Yamada, 1974). In this study, CHASER simulations were conducted at a horizontal resolution of $2.8^{\circ} \times 2.8^{\circ}$, with 36 vertical layers from the surface to ~50 km altitude and a typical time step of 20 min. The meteorological fields simulated by MIROC-AGCM were nudged toward the six-hourly NCEP FNL reanalysis data at every model time step.

The anthropogenic, biomass burning, lightning, and soil emission sources of NO_x were incorporated into CHASER simulations. Anthropogenic emissions were based on the HTAP_v2.2 for 2008. Emissions from biomass burning and soils were inferred based on the ECMWF/MAC reanalysis. The biogenic emissions for VOCs are based on the process-based biogeochemical model the Vegetation Integrative Simulator for Trace gases (VISIT) (Ito and Inatomi, 2012). The NO_x production from lightning is calculated based on the parameterization presented by Price and Rind (1992) linked to the convection scheme of the AGCM (Sudo et al., 2002). The isoprene, terpene, acetone, and ONMV emissions during July were, respectively, 2.14×10^{-11} , 4.43×10^{-12} , 1.60×10^{-12} , and $9.93 \times 10^{-13} \text{ kgCm}^{-2}\text{s}^{-1}$.

Multiple CHASER simulations with different settings used for this study are presented in Table 2.



300 **Table 2 :** The settings of the CHASER simulations used in the current study

Simulation	Anthropogenic emissions	Pyrogenic emissions	Biogenic emissions	Other physical and chemical processes
Standard	ON	ON	ON	ON
L1_HCHO	ON	Pyrogenic VOCs switched OFF	ON	ON
L1_opt	ON	Pyrogenic VOCs switched OFF	Reduced by 50%	ON
L1_NO ₂	ON	Pyrogenic emissions switched OFF	NO ₂ ON	ON
L2	Anthropogenic VOC emissions switched OFF	ON	ON	ON

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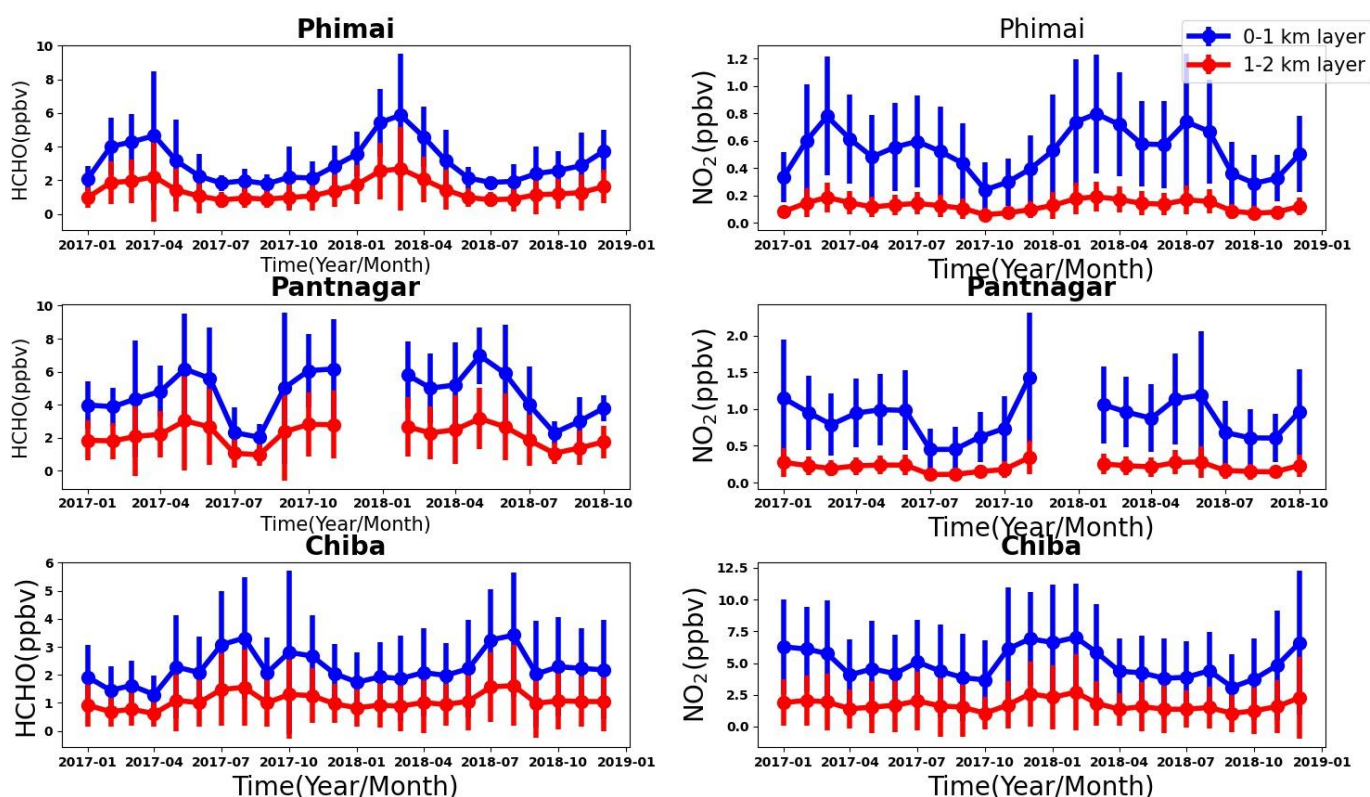


3 Results and discussion

3.1 Results from MAX-DOAS observations

3.1.1 HCHO seasonal variation

The monthly mean HCHO concentrations in the 0–1 and 0–2 km layers from January 2017 – December 2018 and the corresponding one sigma (1σ) standard deviations for the three sites are depicted in Fig. 4. The HCHO amounts at the Phimai site present a consistent seasonal cycle, characterized by high concentrations during the dry season. Such enhancement is related to the influence of biomass burning during the dry season, which has been well documented in the work of Hoque et al. (2018). The HCHO concentrations at Phimai peak in March or April, and the maximum concentrations of 4–6 ppbv. The variation in the peak concentration and timing mainly depend on the intensity of biomass burning activities. During the wet season, the HCHO concentrations are primarily within 2–3 ppbv, indicating a two-fold increase in the HCHO abundances during the dry season. The daily mean HCHO amounts (0–1 km) are 9.84–0.78 ppbv, representing seasonal modulation of 134%.





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320 **Figure 4:** The seasonal variations in the HCHO (left panel) and NO₂ (right panel) concentrations in the 0-1 (blue)
321 and 1-2 (red) km layers in Phimai, Pantnagar, and Chiba. The error bars indicate the one sigma standard deviation
322 of the mean values. The gaps in the plots for the Pantnagar site indicate the unavailability of observations during
323 the investigated period.

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326 The seasonal variation of HCHO in the 0–1 km layer at the Pantnagar site has been discussed by
327 Hoque et al. (2018b). Here, the results are replotted to verify the consistency of the seasonal variations.
328 Observations made during autumn in 2018 were not available because of technical glitches in the
329 spectrometer. Consistent seasonal variation of HCHO abundances is observed at the Pantnagar site, with
330 enhanced concentrations during the spring. The Pantnagar site is affected by biomass burning during
331 spring and autumn (Hoque et al., 2018b), explaining the high concentrations found during spring. In both
332 years, the maximum HCHO concentrations were ~6 ppbv. The springtime peak occurred in May. The
333 HCHO concentrations during the monsoon are ~35% lower than in the spring, indicating a strong effect
334 of the monsoon on the HCHO concentrations found for Pantnagar. The seasonal modulation of HCHO at
335 Pantnagar estimated from the daily mean concentrations is 107%. Under the influence of biomass burning,
336 the maximum monthly HCHO concentrations are similar (~6 ppbv) at Phimai and Pantnagar. The
337 maximum instantaneous concentrations of HCHO during biomass burning influence in Phimai and
338 Pantnagar are 26 and 30 ppbv, respectively. Zarzana et al., (2017) reported HCHO abundances of ~60
339 ppbv in fresh biomass plumes in the US. The lower values obtained from our measurements might be
340 attributable to (1) more aged plumes intercepted by the MAX-DOAS instruments and (2) differences in
341 the types of biomass fuel used. However, comparison of the literature values indicates that our retrieval
342 values of HCHO under biomass burning are reasonable.

343 Higher concentrations characterize the seasonal variation of HCHO at the Chiba site during summer,
344 with a peak ~3 ppbv, likely produced from the oxidation of biogenic VOCs, from which emissions
345 increase with the rising ambient temperature. The HCHO concentrations are ~2 ppbv during another
346 seasons, similar to the HCHO concentrations in Phimai during the wet season. The seasonal modulation



of HCHO in Chiba was ~94%. For a site with similar seasonal variation (i.e., summertime maximum and wintertime minimum), Franco et al. (2015) reported seasonal modulation of HCHO as ~. 88%.

The concentrations of HCHO in the 1–2 km layers at all three sites were lower, almost 50% the value of the concentrations in the 0–1 km layer. The HCHO seasonal modulations at Phimai, Pantnagar, and Chiba sites were, respectively, 131%, 102%, and 90% when calculated based on the HCHO concentration in the 1–2 km layers. The modulation further decreased when retrieved values for 2–3 km layer were used. This finding highlights the strong sensitivity of the MAX-DOAS instrument for near-surface layers, mainly within the planetary boundary layer.

3.1.2 NO₂ seasonal variation at the three sites

Figure 4 also shows the seasonal variation of NO₂ for 0–1 and 1–2 km layers at the three sites. The error bar represents 1 σ standard deviation of the mean values. The NO₂ seasonal variations at Phimai and Pantnagar sites are similar to those of HCHO. A pronounced peak attributable to biomass burning influence is observed during the dry season in Phimai (~0.8 ppbv) and the spring (1.2 ppbv), and the post-monsoon period (1.4 ppbv) in Pantnagar. The lowest NO₂ concentrations in Phimai and Pantnagar were, respectively, ~0.2 and 0.5 ppbv. The NO₂ concentration in Chiba was higher (~7 ppbv) during winter. The longer lifetime of NO_x and lower NO/NO₂ ratio because of lower photochemical activity in winter engender high NO₂ concentrations in Chiba (Irie et al., 2021).

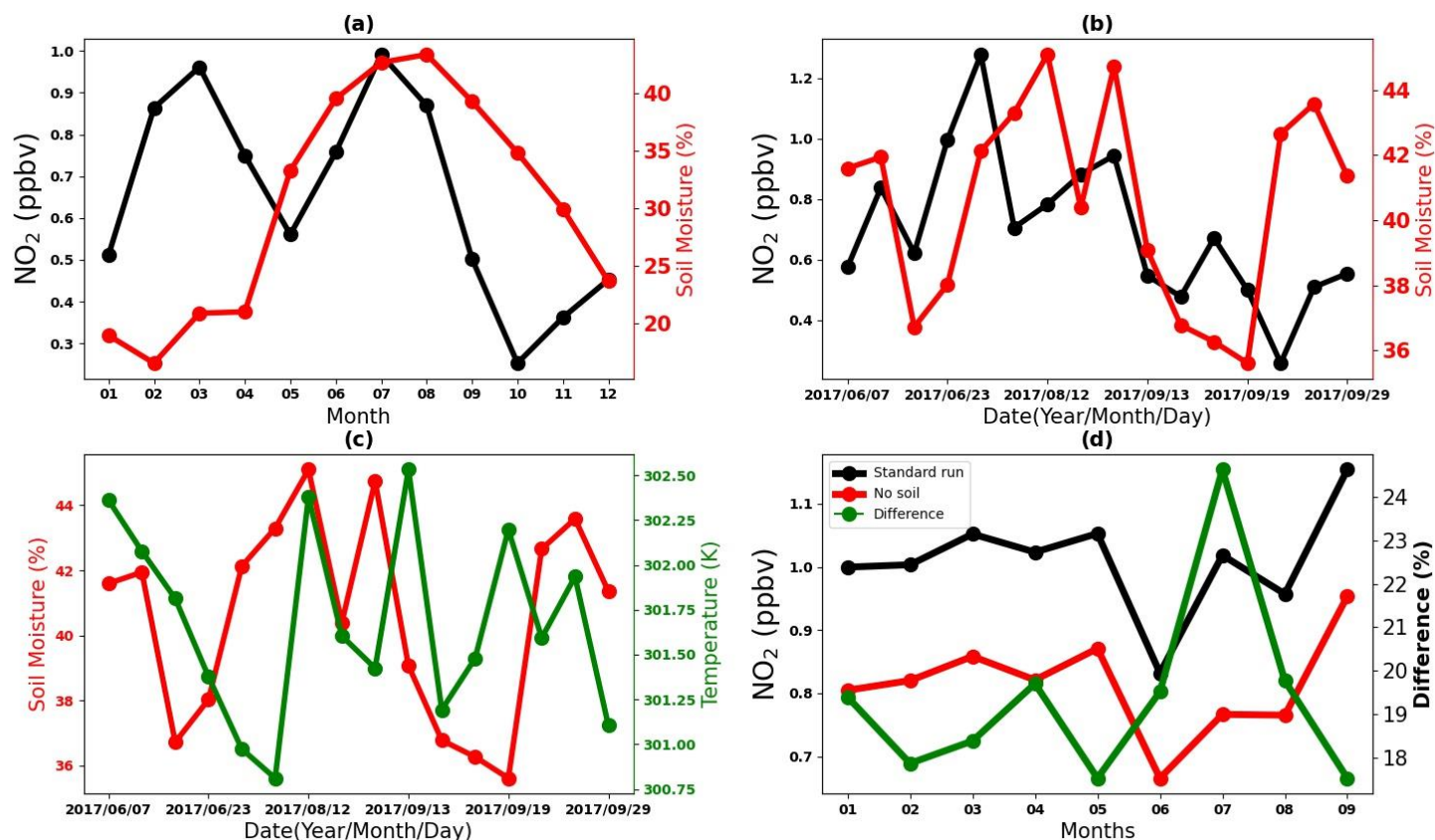
At Phimai, high NO₂ concentrations are observed during the wet season. When Hoque et al. (2018) reported the seasonal variation of NO₂ at Phimai during 2015–2016, such high NO₂ concentrations in the wet season were not observed. It is unlikely that such high concentrations are related to biomass burning. Otherwise, a similar peak would have appeared in the HCHO seasonal variation during the wet season. We also checked the concentrations of glyoxal (CHOCHO) and O₃ (Fig. S1, supplement information) for additional confirmation, simultaneously retrieved using JM2. No enhancement was observed during the wet season. A potential source of NO₂ concentrations during the wet season is NO_x emissions from soil. Although NO₂ is not emitted directly from soils, biological processes in the soil emit NO, which is rapidly converted to NO₂ (Hall et al., 1996). Many studies have established a relation between soil moisture and NO emissions (Carden et al., 1993; Zheng et al., 2000; Schindlbacher et al., 2004; Huber et al., 2020). We use soil moisture data from Soil Moisture Active Passive (SMAP) satellite data to infer the potential



375 contribution of soil NO_x emissions to NO_2 concentrations in Phimai during the wet season. Daily
376 composite soil moisture data (Level 3) with horizontal resolution of 36 km are used for analysis. The data
377 were screened based on the retrieval quality flag, as explained in the user guide document
378 (<https://nsidc.org/data/SPL3SMP/versions/7>, last accessed on 18/05/2021). SMAP observations within
379 the 30 km radius of the Phimai site were selected. Details of the SMAP datasets have been described by
380 Colliander et al. (2017).

381 Figure 5 shows the monthly and daily MAX-DOAS NO_2 concentrations for the 0–1 km layer in
382 Phimai and the SMAP soil moisture and surface temperature data. The coincident dates between the
383 MAX-DOAS and SMAP observations are shown. The expected increase in the soil moisture during the
384 wet season coincided with the high NO_2 concentration in the wet season, indicating more substantial soil
385 NO_x emissions than the dry season. Actually, NO_2 concentrations greater than 0.6 ppbv are primarily
386 associated with soil moisture >40%. Few inconsistencies are observed where the NO_2 concentrations and
387 soil moisture show a reciprocal relation. Reports of many studies have described a negative relation
388 between NO_x emissions and soil moisture (Levine et al., 1990; Davidson et al., 1991; Yamulki et al.,
389 1995). In general, NO_x emissions are more potent at intermediate moisture levels and are comparatively
390 low when the soil is dry or saturated because the additional water fills the pore spaces and restricts NO_x
391 diffusion (Hall et al., 1996). This explanation is consistent with the lower NO_2 concentrations at the end
392 of the wet season despite the higher soil moisture. In addition to soil moisture, the surface temperature is
393 a governing factor for soil NO_x emissions. However, in Phimai, the temperature variation is ~2 K during
394 the wet season, indicating a less significant role in soil NO_x emissions than soil moisture.

395 Because NO_2 emissions from soils are included in CHASER simulations, we attempted to quantify
396 the NO_2 contributions from soil emissions. Figure 5d presents the monthly mean surface NO_2
397 concentrations in Phimai in 2017, simulated including and switching off the soil NO_x emissions. The
398 NO_2 concentrations between 09 and 12 hr were used to calculate the monthly mean concentrations. Soil
399 emissions contribute ~20% of the overall NO_2 concentrations in Phimai, with higher contributions during
400 the wet season. The peak NO_2 concentration (Fig. 3a)) and high soil moisture (Fig. 3b) in July 2017
401 coincide with CHASER sensitivity results, showing the highest contributions (~25%) from soils



402

403 **Figure 5.** (top panel) (a) The seasonal variation of NO₂ concentrations (black colour) in the 0- 1 km layer and the
 404 soil moisture (red colour), retrieved from the MAX-DOAS and SMAP observations, respectively, in Phimai. The
 405 coincident days between the MAX-DOAS and SMAP observations were selected to calculate the seasonal mean
 406 values. (b) The daily mean NO₂ concentrations (black colour) in the 0-1 km layer and the soil moisture (red colour)
 407 are plotted for the wet season's coincident dates. (bottom panel) (c) The surface temperature (green colour) and soil
 408 moisture during the wet season, retrieved from SMAP observations, are plotted. Both parameters are plotted on the
 409 dates when MAX-DOAS observations are available. (d) CHASER simulated NO₂ concentrations in Phimai,
 410 including (black) and switching off (red) the soil NO_x emissions. The green line indicates the percentage difference
 411 between the two simulations.

412

413

414



415

416 occur during July. Consequently, high NO₂ concentrations found during the wet season can be attributed
417 to soil NO_x emissions.

418 The question arises of why such features (high NO₂ concentrations during the wet season) were not
419 found in an earlier study (i.e., Hoque et al., 2018a). The change in soil moisture might explain such a
420 discrepancy. According to our observations, the soil moisture in Phimai is governed by the amount of
421 rainfall. To verify the consistency of the rainfall amounts in Phimai during the wet season, coincident
422 microwave radiometer observations made in 2016 and 2018 were used (Fig. S2, in the supplement
423 information). No significant difference in the precipitable water content was found for those years, which
424 indicates that soil moisture tends to be a less important factor underlying the discrepancy. No discrepancy
425 among our studies is apparent. Our future studies will include detailed discussion of this issue.

426

427 3.1.3 The HCHO to NO₂ ratio (R_{FN}):

428 The HCHO to NO₂ (R_{FN}) ratio is regarded as an indicator of O₃ sensitivity (Martin et al., 2004; Duncan
429 et al., 2010). In this section, the R_{FN} values found for the sites are discussed. To estimate the R_{FN} values,
430 we use the HCHO and NO₂ concentrations in the 0–1 km layer. For discussion of the R_{FN} values, we also
431 used the O₃ concentrations for the 0–1 km layer retrieved from our MAX-DOAS observations. Details of
432 the O₃ retrieval procedure in the JM2 algorithm are explained by Irie et al. (2011). For minimization of
433 stratosphere effects, we used only the O₃ concentrations for SZA < 50°. Irie et al. (2021) reported good
434 agreement between the O₃ concentrations retrieved using JM2 and ozonesonde measurements.

435 Figure 6 shows a scatter plot of daily mean NO₂ and HCHO concentrations, color-coded with the
436 respective O₃ concentrations measured at the respective sites. The O₃ production regime is regarded as
437 VOC-limited for $R_{FN} < 1$ and NO_x limited to when $R_{FN} > 2$ (Duncan et al., 2010; Ryan et al., 2020). This
438 assumption of the R_{FN} values is valid for HCHO concentrations < 10 ppbv (Souri et al., 2020).
439 Consequently, R_{FN} values for HCHO concentrations < 10 ppbv are shown. Almost all high O₃
440 concentrations (>40 ppbv) coincide with the $R_{FN} > 2$ regimes at Phimai and Pantnagar. Only a few, if any,
441 cases were in the $R_{FN} < 1$ regime. At Phimai, few instances during the wet season (O₃ < 20 ppbv) lay in
442 the transition region ($1 < R_{FN} < 2$). Results show that the O₃ production regime is NO_x limited at both sites.



At Chiba, most high O_3 concentration (>80 ppbv) cases are related to high NO_2 concentrations, indicating a VOC-limited regime, which is consistent with results reported by Irie et al. (2021).

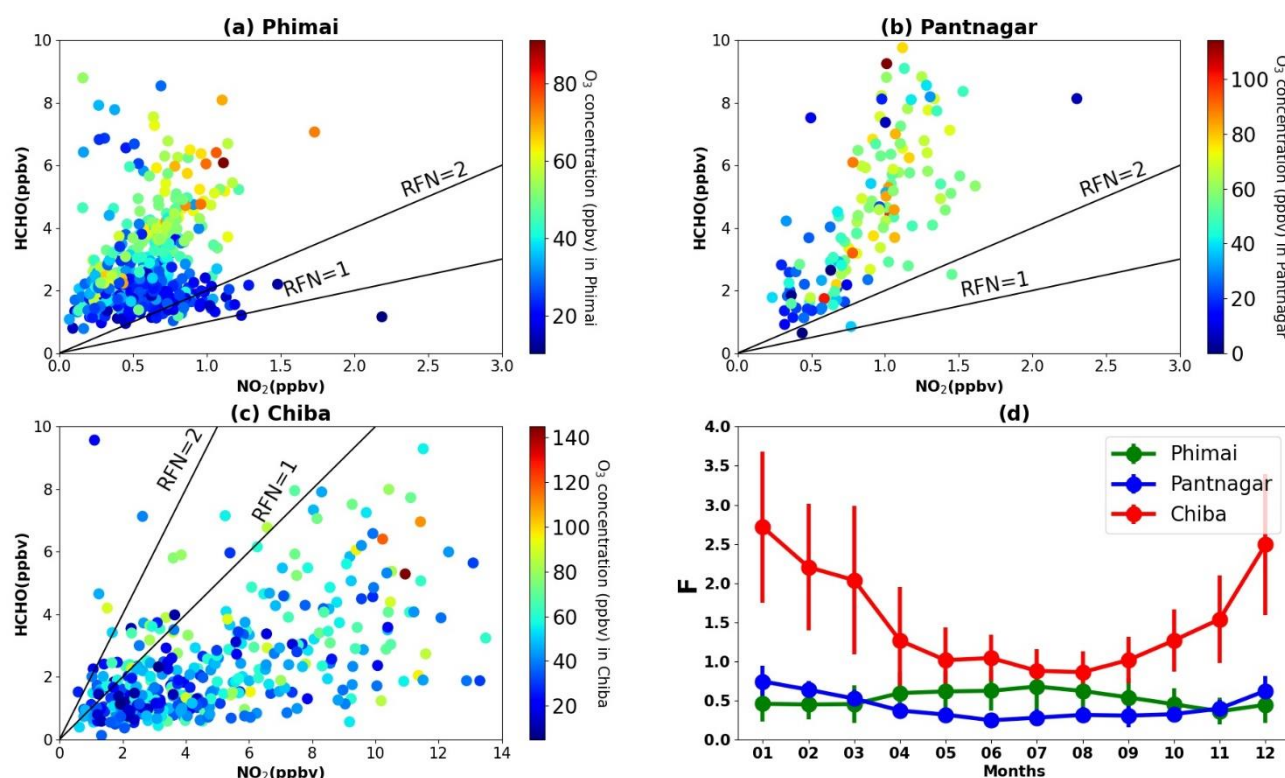


Figure 6. The scatter plot of HCHO and NO_2 concentrations in the 0-1 km layer at (a) Phimai, (b) Pantnagar, and (c) Chiba, coloured with the O_3 concentrations in the 0-1 km layer at the respective sites. The solid lines indicate $R_{FN}=2$ and $R_{FN}=1$ benchmark. (d) The seasonal variations in the column to surface conversion factor (F) for the Phimai, Pantnagar, and Chiba sites, estimated from the CHASER simulated HCHO and NO_2 surface concentrations and vertical column density. The simulated data from 07:00 – 18:00 in 2017 were used to estimate the F values. The error bars indicate the one sigma standard deviation of the mean values.

Mahajan et al. (2015) reported OMI-derived R_{FN} values < 1 over the IGP region, which contradicts our findings for Pantnagar. However, the R_{FN} values at Pantnagar are consistent with the values reported by Biswas et al. (2019), derived from MAX-DOAS observation at another site in the IGP region. The O_3



sensitivity inferred from the R_{FN} values estimated from satellite and surface in-situ measurements showed different results because of the gradient of the R_{FN} values. Jin et al. (2017) proposed the idea of a conversion factor to account for gradient differences in the surface and column-derived R_{FN} values, estimating the conversion factor from the model simulated surface and column abundances of NO_2 and HCHO. For this study, we adopt the method reported by Jin et al. (2017) using the CHASER simulated NO_2 and HCHO concentrations and vertical columns.

First, the CHASER simulated near-surface NO_2 and HCHO concentrations were converted to number density and the effective boundary layer height (E) (Halla et al., 2011; Jin et al., 2017) was estimated.

$$E_{\text{NO}_2} = \frac{\text{NO}_2 \text{ total coulumn}}{\text{NO}_2 \text{ near-surface number density}} \quad (6)$$

$$E_{\text{HCHO}} = \frac{\text{HCHO total coulumn}}{\text{HCHO near-surface number density}} \quad (7)$$

Therein, E_{NO_2} and E_{HCHO} respectively denote the effective boundary layer heights of NO_2 and HCHO.

In the second step, the column to surface conversion factor (F) was calculated according to the following equation:

$$F = \frac{E_{\text{HCHO}}}{E_{\text{NO}_2}} \quad (8)$$

The seasonal variation of F for the three A-SKY sites and the associated 1σ standard deviation of the mean values are depicted in Fig. 6d. The F values over East Asia reported by Jin et al. (2017) were ~ 2 , with no marked seasonal variation. Compared to values reported in the literature, the CHASER estimated F values over Chiba range between 1–2.5, which seems reasonable. The reason for the discrepancy in the F values can be (1) the difference in the model resolutions between the literature and current study, (2) the literature values are reported for polluted regions ($\text{NO}_2 > 2.5 \text{ molecules cm}^{-2}$) and considered the simulation data for 1–2 PM. By contrast, the estimates of this study included daytime (07:00 – 18:00) values.



481 The F values in Pantnagar are mostly < 1 , with no distinctive seasonal variation. When this estimated
482 conversion factor is used with the values of Mahajan et al. (2015) (i.e., OMI derived R_{FN} values < 1 over
483 the IGP region), the discrepancy in the satellite and ground-based observation derived R_{FN} values in the
484 IGP region are reduced, indicating that the estimated F values for the Pantnagar site can be representative
485 for the IGP region. The F values at the Phimai site range were 0.5–1. No report of the relevant literature
486 presents F values for the southern and southeastern Asian regions. Consequently, our estimated F values
487 for the Phimai and Pantnagar site are useful as representative values for these respective regions, which
488 can be improved further based on the results.

489

490 **3.1.4 Can the glyoxal to formaldehyde (R_{GF}) ratio be used as an indicator of ozone sensitivity?**

491 Ryan et al. (2020) proposed that, in addition to the R_{FN} , the glyoxal to formaldehyde ratio (R_{GF}) can also
492 be a helpful indicator of O_3 production regimes and concentrations because VOCs are an important
493 component in the VOC- HO_x - NO_x catalytical cycle governing O_3 production. We use the coincident
494 glyoxal (CHOCHO) concentrations retrieved using the JM2 algorithm and the O_3 concentrations to test
495 this hypothesis. Detailed procedures and error estimates of the CHOCHO retrievals are explained by Irie
496 et al. (2011) and by Hoque et al. (2018a). The R_{GF} values estimated from HCHO and CHOCHO retrieved
497 using the JM2 algorithm have been discussed by Hoque et al. (2018a, 2018b). Figure 7 shows a scatter
498 plot of the daily mean CHOCHO and HCHO concentrations, colored with the O_3 concentrations for the
499 three sites. The R_{GF} values are < 0.04 when VOC sources are dominantly anthropogenic or pyrogenic,
500 whereas higher R_{GF} values are observed under the dominant biogenic sources of VOCs (Hoque et al.,
501 2018a; Vrekoussis et al., 2009). At the three sites, high O_3 concentrations (>40 ppbv) are associated
502 primarily with $R_{GF} < 0.04$, consistent with findings reported by Ryan et al. (2020). However, for O_3
503 concentrations below 20 ppbv, the robustness of the R_{GF} decreases. At all three sites, many cases are
504 found when the lower O_3 concentrations are associated with R_{GF} values less than 0.04. Moreover, two
505 issues should be accounted for before using R_{GF} as an indicator of O_3 production sensitivity. (1) The
506 change of R_{GF} under different VOC emission scenarios is still under discussion. There are contrasting
507 results from field observations (Digangi et al., 2012; Kaiser et al., 2015; etc.). (2) Also, the gradient of
508 the R_{GF} values is not well understood because few studies have particularly addressed the issue. Therefore,



we infer that a more detailed analysis is necessary to regard R_{GF} as an indicator of O_3 sensitivity, in addition to R_{FN} .

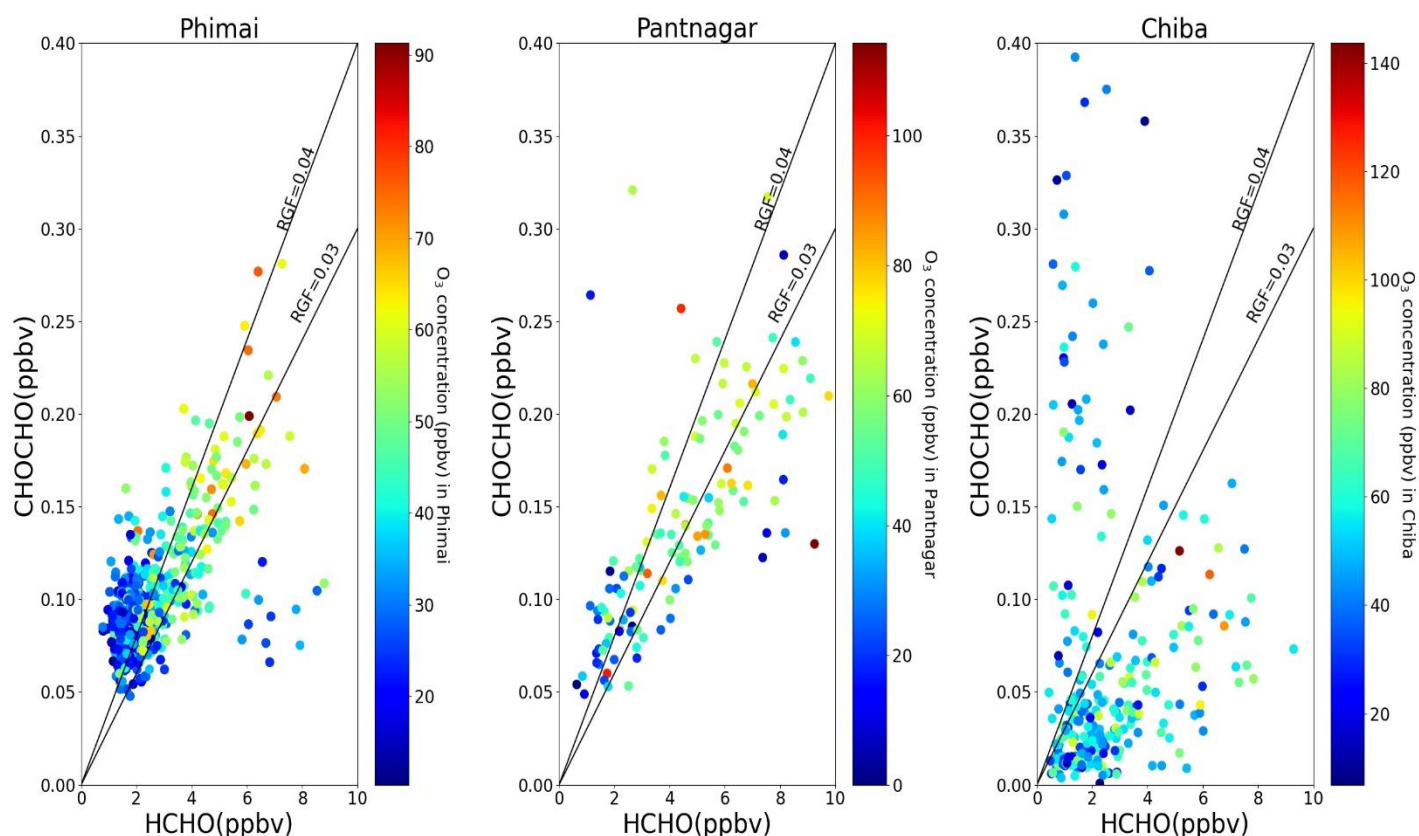


Figure 7. The scatter plots of CHOCHO and HCHO concentrations in the 0-1 km layer at Phimai, Pantnagar, and Chiba, coloured with the O_3 concentrations in the 0-1 km layer at the respective sites. The solid lines indicate the R_{GF} values at 0.04 and 0.03, respectively.



522 3.2 MAX-DOAS and CHASER comparison

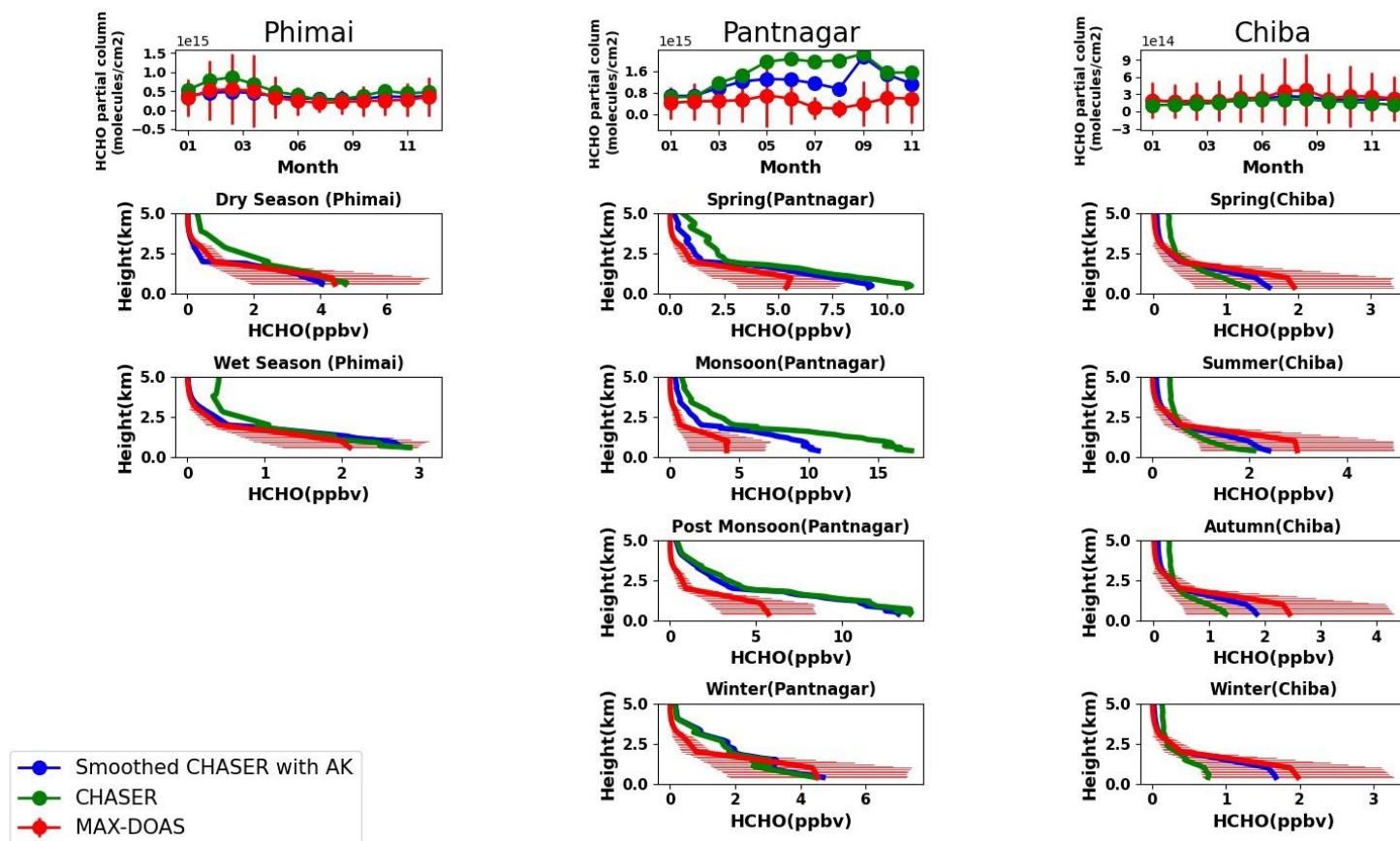
523 3.2.1 Comparison between MAX-DOAS and CHASER HCHO

524 Figure 8 shows seasonal averages of the HCHO profiles and partial columns (1–4 km) at the three
525 SKYNET sites for the MAX-DOAS and CHASER datasets. In addition, the CHASER outputs smoothed
526 with MAX-DOAS averaging kernels (AK) are depicted. The AK was applied following the procedure
527 described by Franco et al. (2015). First, the CHASER HCHO profiles were interpolated to the MAX-
528 DOAS vertical grids. Then the averaging AK was applied to the daily mean interpolated profiles. Next,
529 the MAX-DOAS AK information from individual retrieved profiles was seasonally averaged according
530 to the climate classifications of each site. Finally, the CHASER output for the days when observations
531 are available was selected for the comparison. The numbers of days available for comparison at Phimai,
532 Pantnagar, and Chiba were, respectively, 690, 340, and 668.

533 At the Phimai and Chiba sites, the HCHO partial columns of both datasets agree well within the 1σ
534 standard deviation of the MAX-DOAS observations. The HCHO partial column's enhancement because
535 of biomass burning in Phimai during the dry season was captured by CHASER. The HCHO profiles of
536 both datasets also show good agreement at the Chiba and Phimai sites. In both datasets, the peak
537 concentration of HCHO during the dry season in Phimai is ~ 5 ppbv, indicating that the biomass burning
538 estimates used in CHASER are reasonable. At both sites, CHASER has exhibited good skill at
539 reproducing HCHO concentrations below 2 km. Actually, CHASER simulates slightly lower HCHO
540 concentrations than MAX-DOAS at the Chiba site and overestimates HCHO during the wet season in
541 Phimai. However, CHASER HCHO profiles are mainly within the 1σ standard deviation range of the
542 ground-based observations. Application of AK to the CHASER profiles improves agreement between the
543 datasets, as Table 3 shows. At both sites, however, larger discrepancies are observed above 2 km. Such
544 differences are expected because of the reduced sensitivity of MAX-DOAS in the upper tropospheric
545 layers.



546



547

548 **Figure 8.** The seasonal variations in the HCHO partial columns from 1 to 4 km and vertical profiles during all
 549 seasons at the three A-SKY sites, inferred from the MAX-DOAS observations (red) and CHASER
 550 simulation (green). The CHASER HCHO partial column and vertical profile smoothed with the MAX-DOAS AK
 551 are shown in blue colour. The AK information of all the screened (as explained in section 2.2) retrievals was
 552 averaged based on the respective sites' seasonal classification. The selected CHASER simulated time and date
 553 corresponds to the available MAX-DOAS observations at the individual sites. The error bars indicate the one sigma
 554 standard deviation of mean values of the MAX-DOAS observations.

555

556

557



Table 3: The seasonal differences between the observed and modelled HCHO concentrations in the 0-1 km layer in Phimai and Chiba. The units of the differences are given in ppbv.

Site	Season	MAX-DOAS – CHASER	MAX-DOAS – smoothed CHASER
Phimai	Dry	-0.34	-0.12
Phimai	Wet	-0.68	-0.31
Chiba	Spring	0.96	0.71
Chiba	Summer	-0.02	-0.10
Chiba	Autumn	-0.38	-0.22
Chiba	Winter	0.27	-0.05

560

561

At the Pantnagar site, considerable discrepancies were found between data of CHASER and MAX-DOAS, mainly during the monsoon and post-monsoon seasons. During the monsoon season, CHASER-simulated HCHO concentrations show a peak. The comparison results did not improve despite application of AK to the CHASER output. In most cases, the CHASER simulated values are almost twice the observation values. To infer the potential reason for the discrepancies, the annual mean HCHO concentrations at all three sites are shown in Figure 9. CHASER-simulated HCHO concentrations at Phimai and Pantnagar are similar, consistent with the MAX-DOAS observations at both locations (Fig. 1). Consequently, emission estimates used in the CHASER simulations are unlikely to contribute to the marked discrepancies found for the comparison at Pantnagar. Mountainous terrain exists a few kilometers away from the Pantnagar site. The differences can originate from treatment of the complex landscapes in



the retrieval and simulations. The radiative transfer calculations for MAX-DOAS retrieval consider a flat terrain. Such an assumption is expected to have a non-significant effect on retrieval because the horizontal resolution of the MAX-DOAS system is ~ 10 km.

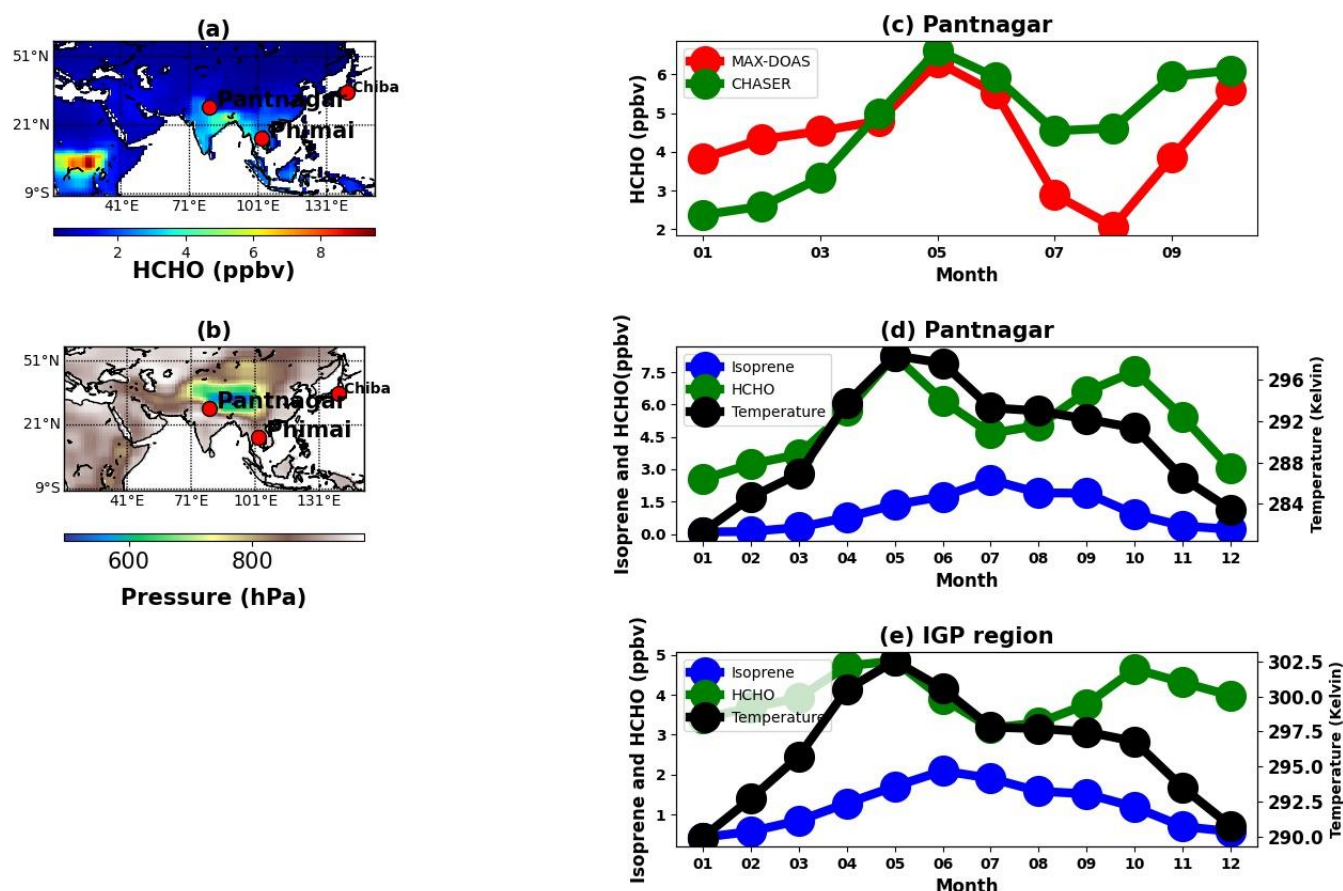


Figure 9. (left panel) (a) The mean surface HCHO concentrations at the three SKYNET sites in 2017, simulated by CHASER. (b) The simulated mean surface pressure at the respective SKYNET sites. (right panel) (c) The seasonal variations in the MAX-DOAS (red) and CHASER (green) HCHO concentrations in Pantnagar. The coincident dates between the observations and model are plotted only. The CHASER simulated seasonal variations in HCHO (green), isoprene (blue), and temperature (black) in 2017 in (d) Pantnagar and (e) the IGP region. Only the daytime simulated values were used for the plots (a), (b), (d), and (e).



584

585 Moreover, the MAX-DOAS HCHO products in Pantnagar showed good agreement with satellite
586 observations (Hoque et al., 2018b). Figure 9b shows the CHASER simulated surface pressure results
587 obtained for all three sites. The lower surface pressure at Pantnagar, compared to the other sites, indicates
588 that the surface elevation of Pantnagar is higher because of the mountains around the site. The simulated
589 surface height at Pantnagar is ~1700 m, although it is ~300 m in Phimai and Chiba. Consequently, when
590 CHASER values at Pantnagar are extrapolated to the MAX-DOAS grid (~100 m), the surface HCHO
591 concentrations are considerably different from the observations. The CHASER HCHO simulation at
592 Pantnagar is compared directly with the MAX-DOAS observations, as depicted in Fig. 9(c). Only the
593 coincident days between the observation and model are shown.

594 The HCHO seasonal variations in Pantnagar during the winter and spring are well reproduced by
595 CHASER. The peak HCHO concentrations during spring are ~6 ppbv in both datasets. The most marked
596 discrepancies occurred during the monsoon season, during which CHASER values are largely
597 overestimated. To infer the potential reason for the discrepancy between the model and observation,
598 CHASER simulated isoprene concentrations, temperature, and HCHO concentrations at the Pantnagar
599 site, as depicted in Fig. 9(d).

600 Oxidization of precursor hydrocarbon and photochemical reactions are the most dominant sources of
601 HCHO. Isoprene is the most abundant hydrocarbon in the atmosphere. The average ambient isoprene
602 concentrations during July, August, and September in the IGP region are 1.4 ± 0.3 ppbv (Mishra et al.,
603 2020). Therefore, a CHASER isoprene concentration range of 3–4 ppbv during the monsoon season seems
604 reasonable. The high concentrations of HCHO during the monsoon season coincide with the high isoprene
605 concentrations and temperature in Pantnagar. The tight correlation between the isoprene and HCHO
606 concentrations suggests that the CHASER simulated HCHO at Pantnagar is governed by the isoprene
607 seasonal modulation. Also, MAX-DOAS HCHO concentrations are lower during the monsoon season,
608 likely because of the strong wet deposition and suppressed photochemical activities. Consequently, the
609 physical processes driving the HCHO concentration at Pantnagar in the second half of a year differ from
610 those considered in the model.



611 Surl et al. (2018) found good agreement between OMI and GEOS-CHEM HCHO columns in the IGP
612 region only during the first half of the year. The HCHO, isoprene, and temperature variations in the whole
613 IGP region are shown in Fig. 9(e). Reportedly, the HCHO concentrations in the IGP region reach a peak
614 during the spring and post-monsoon seasons. Strong correlation between HCHO, isoprene, and
615 temperature variation in the first half of the year indicates that the change in biogenic emissions strongly
616 drives the HCHO seasonal modulation. During the post-monsoon period, it is likely that biomass burning
617 and anthropogenic emissions are more dominant than biogenic emissions because HCHO modulation
618 differs from those of isoprene and temperature. CHASER simulations show low HCHO concentrations
619 in the IGP region during the monsoon despite high isoprene concentrations and temperature. The HCHO
620 seasonal variation in the IGP is similar to that reported by Mahajan et al. (2015), which indicates that the
621 physical processes driving the HCHO seasonality in the IGP region are well reflected in the CHASER
622 simulations. Consequently, the assumptions of CHASER simulations (i.e., emissions and physical
623 processes) are unlikely to be an important reason for the discrepancy in the comparison for Pantnagar.
624 Because CHASER demonstrated good capability to reproduce the HCHO seasonal variation at a regional
625 scale (i.e., in the IGP region), the discrepancies at the local scale comparison (i.e., in Pantnagar) are
626 expected to decrease with high resolution (improved spatial resolution) simulations.

627

628 3.2.2 Comparison between MAX-DOAS and CHASER NO₂

629 Figure 10 presents the seasonal averages of the NO₂ profiles and partial columns (<4 km) at the Phimai
630 and Chiba sites for the MAX-DOAS and CHASER datasets. Calculation of the smoothed CHASER
631 output is similar to that explained in the section 3.2.2. Data of comparisons in Phimai and Chiba are
632 presented in Table 4. The smoothed CHASER NO₂ partial column and profiles show good agreement
633 with the MAX-DOAS datasets at both locations. CHASER well reproduced the enhanced NO₂ during the
634 dry season in Phimai. Amidst the biomass burning influence, the NO₂ concentrations in Phimai are mostly
635 < 1 ppbv. Consequently, CHASER demonstrated good skills in a region characterized by low NO₂
636 concentrations (<1 ppbv). CHASER simulations do not show the NO₂ enhancement related to soil NO_x
637 emissions during the wet season. However, sensitivity studies with CHASER simulations quantified the
638 soil NO_x contribution to the NO₂ concentrations in Phimai (Fig. 5). This result indicates that CHASER



captured the soil NO_x contribution, but the magnitude differs because of the horizontal resolution of the datasets.

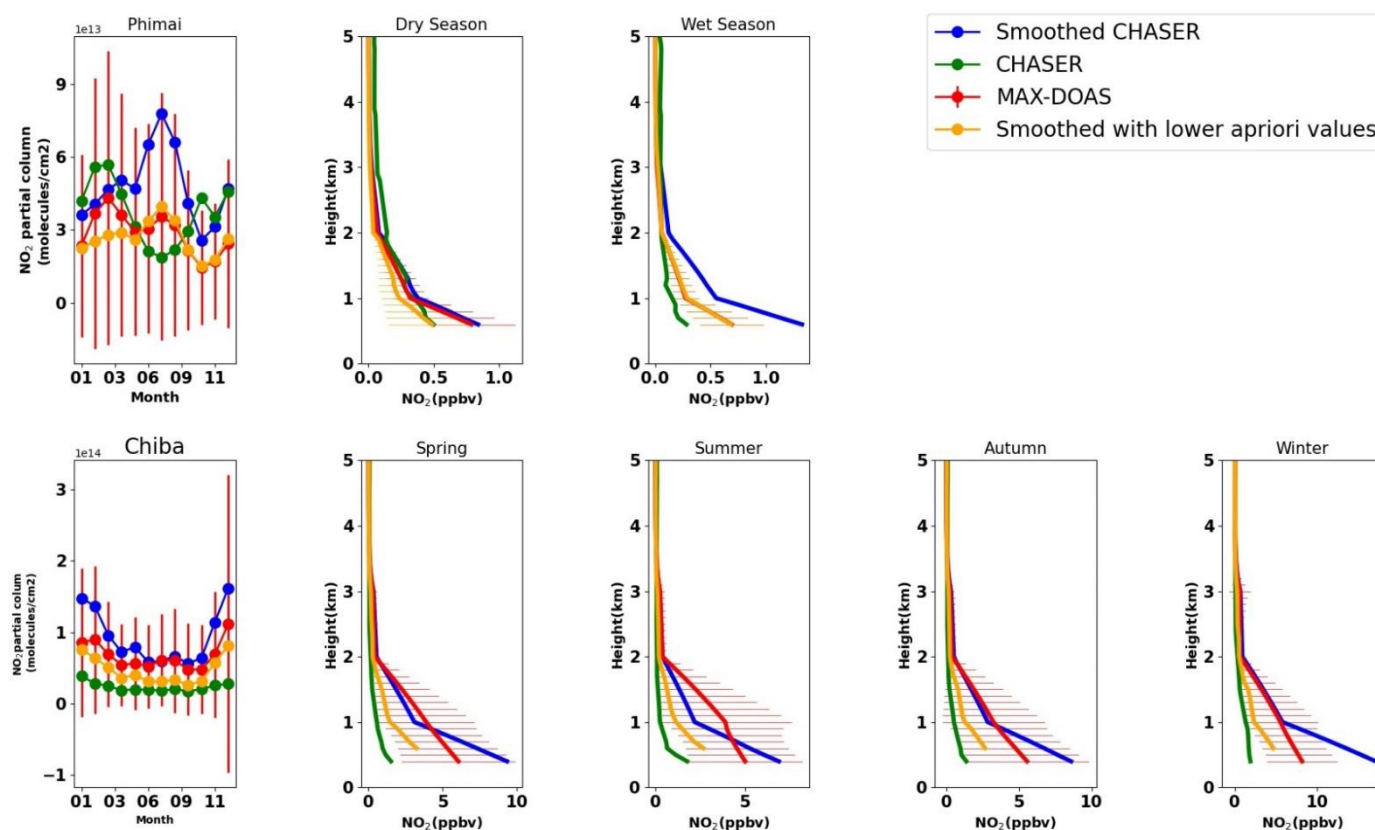


Figure 10. The seasonal variations in the NO_2 partial columns (< 4 km) and vertical profiles during all seasons at Phimai and Chiba, inferred from the MAX-DOAS observations (red) and CHASER simulation (green). The CHASER NO_2 partial columns and vertical profiles smoothed with the MAX-DOAS AK are shown in blue color. The orange colour indicates the smoothed CHASER profile when calculated with a priori values reduced by 50%. The AK information of all the screened (as explained in section 2.2) retrievals was averaged based on the respective sites' seasonal classification. The selected CHASER simulated time and date corresponds to the available MAX-DOAS observations at the respective sites. The error bars indicate the one sigma standard deviation of mean values of the MAX-DOAS observations.



652

653 **Table 4:** The seasonal differences between the observed and modelled NO₂ concentrations in the 0-1 km
 654 layer in Phimai and Chiba. The unit of the differences are in ppbv.

Site	Season	MAX-DOAS – CHASER	MAX-DOAS – smoothed CHASER
Phimai	Dry	-0.12	-0.17
Phimai	Wet	-0.02	-0.70
Chiba	Spring	2.69	-0.91
Chiba	Summer	2.70	-0.30
Chiba	Autumn	2.21	-1.02
Chiba	Winter	2.21	-1.71

655

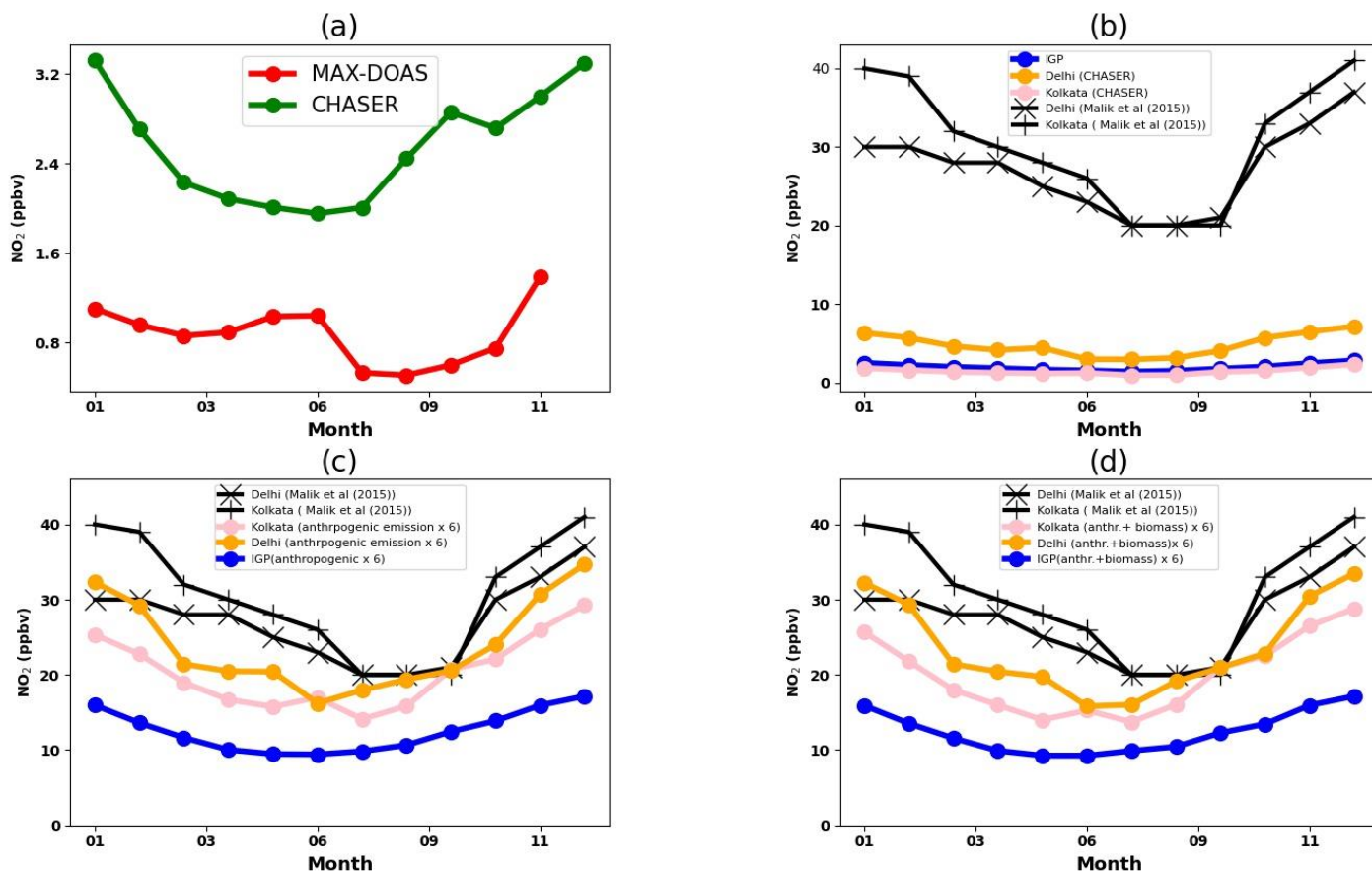
656

657 Results show that CHASER underestimated the NO₂ concentration in Chiba, although the smoothed
 658 CHASER output agrees well with observations. The simulated values were mainly within one sigma
 659 standard deviation of the MAX-DOAS values. At both sites, smoothed NO₂ profiles at 0–2 km were in
 660 good agreement with observations. Although the smoothed output improved the comparison results, the
 661 variation in the smoothed output differed from the model output, mostly during all seasons in Chiba and
 662 during the wet season in Phimai. Such differences are likely to be related to the apriori data used for
 663 smoothing the CHASER output. Figure 10 includes the smoothed output produced when a priori values
 664 are decreased by 50%. The a priori data used to retrieve the profile information are taken from the
 665 retrieved VCD and SCD values. Because MAX-DOAS observations are most sensitive to the lower



666 tropospheric layers, the apriori values affected the smoothed model output, mostly within the 0–2 km
667 layers. Therefore, the discrepancies between CHASER and smoothed CHASER profiles are lessened
668 when lower apriori values are used. The overestimated values in the smoothed output during the wet
669 season in Phimai and spring and winter in Chiba were also improved when lower apriori values were
670 used. Although the apriori values primarily affect the smoothed profile shapes, that variation is mostly
671 within 1σ standard deviation of the observations. At Phimai, the model output shows better agreement
672 with observations than that found for the smoothed outputs. At a low NO_2 concentration concentration ($<$
673 1 ppbv), the AK information seems to be less significant if the model can reproduce low concentration
674 scenarios.

675 Based on the discussion of HCHO comparisons in Pantnagar, CHASER-simulated NO_2
676 concentrations at Pantnagar are compared directly with observations depicted in Fig. 11. The NO_2
677 seasonal variation in the datasets disagrees considerably. To infer the potential reasons for the
678 disagreement, the NO_2 simulations in the whole IGP region and the simulations around two megacities in
679 the IGP region, New Delhi and Kolkata, are shown in the figure. In addition, NO_2 surface measurements
680 in the megacities reported by Malik and Lal (2015) are considered. The lowest concentrations during the
681 monsoon characterize the overall NO_2 variation in the IGP. The CHASER NO_2 simulations show similar
682 seasonal modulation in both cities. The observed and modeled NO_2 seasonal variability in New Delhi and
683 Kolkata agree well with R values of 0.80 despite the underestimated model values. To improve the
684 agreement in terms of absolute values, simulations with optimized anthropogenic emissions are conducted
685 and presented in Figs. 11(c) and 11(d). Increasing the anthropogenic emissions improves the agreement
686 between the model and observations, mostly in New Delhi. On a seasonal scale, the maximum change in
687 the NO_2 concentration in the megacities are between 10–20 and 15–25 ppbv, according respectively to
688 the observations and model. Consequently, despite underestimation, the CHASER NO_2 simulations seem
689 reasonable.



690

691

692 **Figure 11.** (a) The seasonal variation in the NO₂ concentrations in the 0-1 km layer in Pantnagar was obtained
 693 from the MAX-DOAS observations (red) and CHASER simulations (green). The coincident dates between the
 694 datasets were selected for the plot. (b) The CHASER simulated surface NO₂ concentrations in New Delhi, Kolkata,
 695 and the IGP region. New Delhi and Kolkata are two megacities located in the IGP region. The black curves show
 696 the surface NO₂ concentrations in New Delhi and Kolkata reported by Malik and Lal (2015). The CHASER
 697 simulated NO₂ concentrations in the IGP region, New Delhi, and Kolkata (c) with increased anthropogenic
 698 emissions and (d) with increased anthropogenic and biomass burning emissions. The observational data in both (c)
 699 and (d) are taken from the work of Malik and Lal (2015)

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703 Similar results are obtained when biomass burning emissions are optimized and anthropogenic
704 emissions (Fig. 11(d)) are included, indicating that the anthropogenic emissions' changes primarily drive
705 the NO₂ variation found in the IGP region. In contrast, the NO₂ seasonality in Pantnagar is influenced
706 strongly by biomass burning (according to observations), thereby leading to the discrepancy between the
707 observations and model in Pantnagar. In the case of Pantnagar, the CHASER shows better skills at
708 reproducing the climatology of NO₂ and HCHO on a regional scale.

709

710 **3.2.3 Contribution from biomass burning to the HCHO and NO₂ abundances in Phimai and** 711 **Pantnagar**

712 This section presents estimation of the contribution of biomass to the HCHO and NO₂ concentrations in
713 Phimai and Pantnagar. Figure 12 presents results of simulation L1_HCHO, L1_opt, and L1_NO₂. The
714 simulation settings are shown in Table 2. For better readability, the switched-off emissions criterion is
715 described in the legends of Fig. 12. All the plots present the mean concentrations in the 0–1 km layer.
716 Biomass burning contributes ~10% to the HCHO concentrations in Phimai during the dry season.
717 However, based on the observations, a more substantial effect of biomass burning is expected. During the
718 wet season, the MAX-DOAS and CHASER HCHO concentrations are, respectively, ~2 and ~4 ppbv,
719 indicating overestimation of the biogenic emissions in CHASER. Figure 12(b) shows the HCHO
720 concentration obtained from simulation L1_opt and MAX-DOAS observations in 2017. In the L1_opt
721 simulation setting, the biomass burning emissions were switched off; the biogenic emissions were
722 optimized to reproduce results analogous to those obtained from observations made during the wet
723 season. In the absence of biomass burning, the surface HCHO concentrations in Phimai would be ~2 ppbv,
724 resulting in a biomass burning contribution of ~20–50% during the dry season. The observed interseason
725 difference in the HCHO concentration in Phimai is ~60%. Consequently, the revised biomass burning
726 contribution estimate is apparently more reasonable. Pyrogenic emissions contributions to the NO₂
727 concentrations in Phimai are ~10% during the dry season (Fig. 12(c)). Because the NO₂ concentrations
728 are low in Phimai, the simulation during March, when the influence of biomass burning is highest, is used
729 to derive a better contribution estimate. As depicted in Fig. 12(d), in the absence of biomass burning, the



NO₂ concentration during March would be ~0.84 ppbv, indicating a contribution up to 35% to the NO₂ concentrations in Phimai.

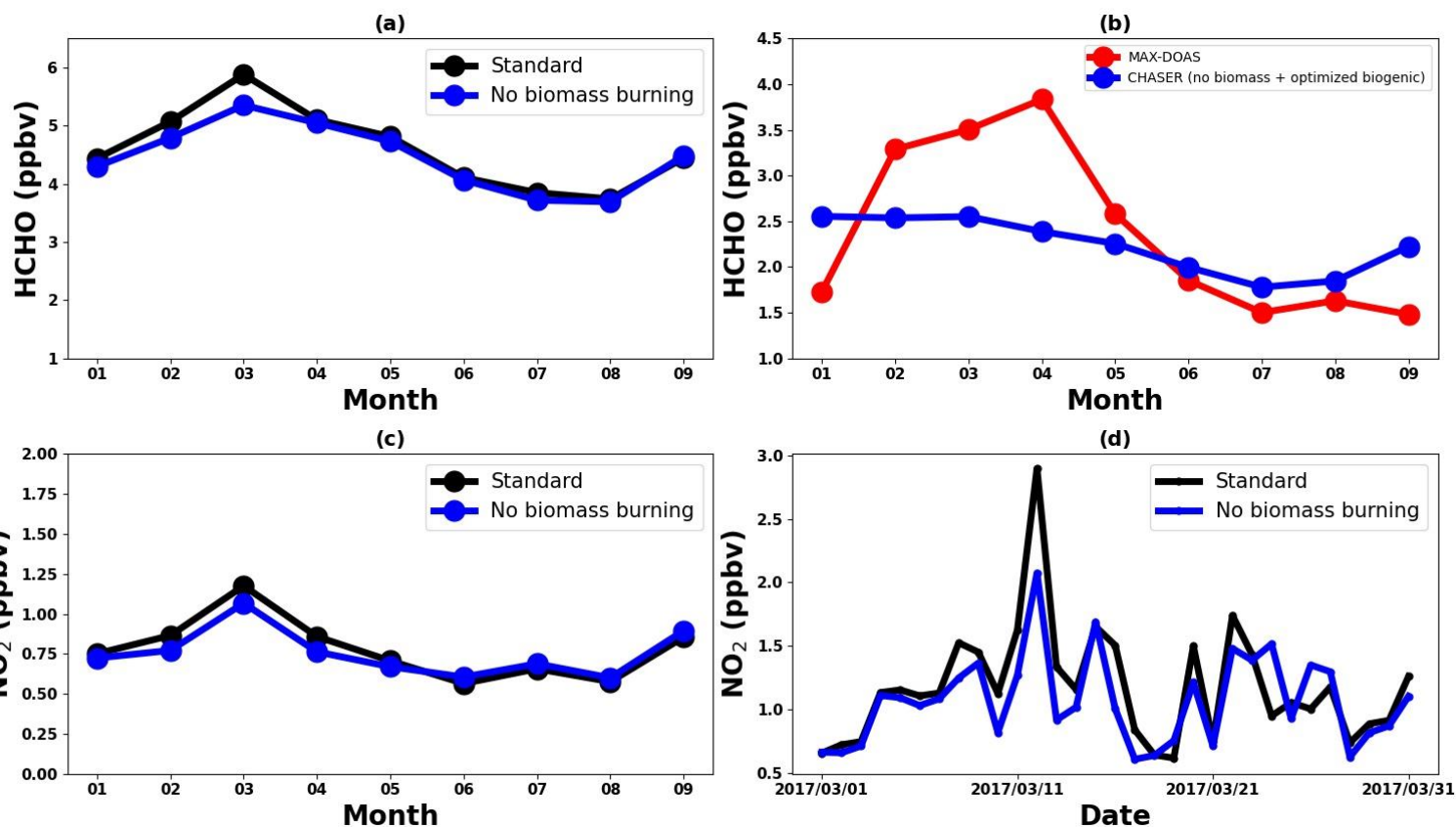


Figure 12. (top panel) The seasonal variations of the HCHO concentrations in the 0-1 km layer in Phimai, obtained from the standard and L1_HCHO simulations. The pyrogenic emissions of VOCs are switched off in the L1_HCHO simulations. (b) The HCHO seasonal variability in Phimai in 2017 was obtained from the MAX-DOAS observations (red) and L1_opt simulations. In addition to switching off the pyrogenic VOC emissions, the biogenic emissions were reduced by 50% in the L_opt simulation. The coincident dates between the observation and the simulations are plotted only. (bottom panel) (c) The seasonal variation of NO₂ in Phimai was obtained from the standard and L1_NO₂ simulations. The pyrogenic NO₂ emissions were switched off in the L1_NO₂ simulations. (d) The daily NO₂ concentrations in Phimai during March, obtained from the standard and L1_NO₂ simulations.



742 All the simulations are plotted for the year 2017, and only the daytime values from 09:00 – 15:00 LT are used for
743 calculating the seasonal mean.

744

745

746 At Pantnagar, pyrogenic emissions contribute up to 13% to the HCHO concentrations, which is
747 apparently underestimated based on the observed seasonality (Fig. 13(a)). Because CHASER showed
748 better capability on a regional scale, the effect of pyrogenic emissions on the HCHO variation in the IGP
749 region is considered. According to L1_HCHO simulation results, the effect of biomass burning emissions
750 on the regional HCHO modulation is also apparently non-significant. The HCHO concentrations in India
751 have biogenic, anthropogenic, and pyrogenic VOC sources. Biogenic VOCs are the primary driver of the
752 over HCHO variation (Surl et al., 2018). Consequently, two potential reasons might be responsible for
753 small effects of pyrogenic emissions on HCHO concentrations: (1) Overestimation of the biogenic
754 emission or underestimation of pyrogenic emissions in the model. (2) Stronger effects of anthropogenic
755 VOC emissions than those of pyrogenic VOCs. In fact, CHASER overestimated the HCHO
756 concentrations at both sites, indicating overestimation of biogenic emissions. To infer the role of
757 anthropogenic emissions on the HCHO concentration, the results of simulation L2 are depicted in Figs.
758 13(a) and 13(b). Anthropogenic emissions contribute up to 30% of the HCHO concentration in the IGP
759 region, with a maximum contributed during the post-monsoon season. The greatest effect of
760 anthropogenic emissions during the post-monsoon seasons coincides with the lower isoprene
761 concentration (i.e., biogenic emissions) and temperature (Fig. 9(e)). Consequently, anthropogenic
762 emissions are likely to be a significant driver of HCHO concentrations in the IGP region after biogenic
763 emissions. At Pantnagar, anthropogenic VOCs contribute more (~. 30%) than pyrogenic emissions do.

764 It is particularly interesting that during the spring, CHASER simulations show that biomass burning
765 contributes up to 8% and 15% to the NO₂ concentrations, respectively, in Pantnagar and the IGP region
766 (Fig. 13(c)). According to results reported by Hoque et al. (2018b), biomass burning is expected to have
767 a stronger influence on NO₂ and HCHO climatology in Pantnagar. A reasonable estimate of the
768 contributions from different emission sources in Pantnagar can likely be obtainable from higher resolution
769 CHASER simulations. However, the regional contribution estimates are reasonable.



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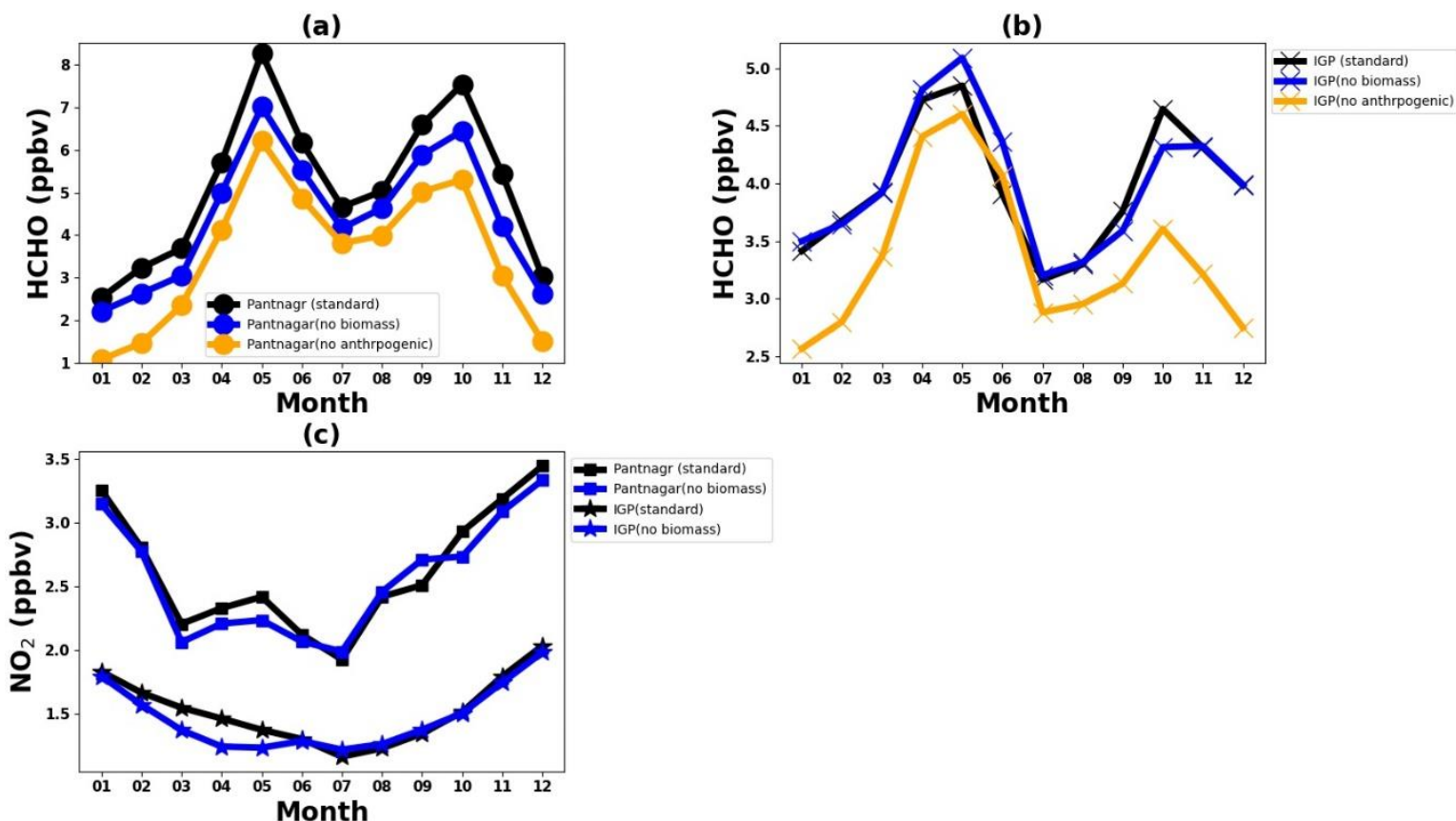


Figure 13. The seasonal variations of the HCHO concentrations in the 0-1 km layer in (a)Pantnagar and (b) in the IGP region, obtained from the standard, L1_HCHO, and L2 simulations. VOCs' pyrogenic emissions and anthropogenic emissions are switched off in the L1_HCHO and L2 simulations, respectively. (c) The seasonal variation of NO₂ in Pantnagar and the IGP region was obtained from the standard and L1_NO₂ simulations. The pyrogenic NO₂ emissions were switched off in the L1_NO₂ simulations. All the simulations are plotted for the year 2017. All the simulations are plotted for the year 2017, and only the daytime values from 09:00 – 15:00 LT are used for calculating the seasonal mean.

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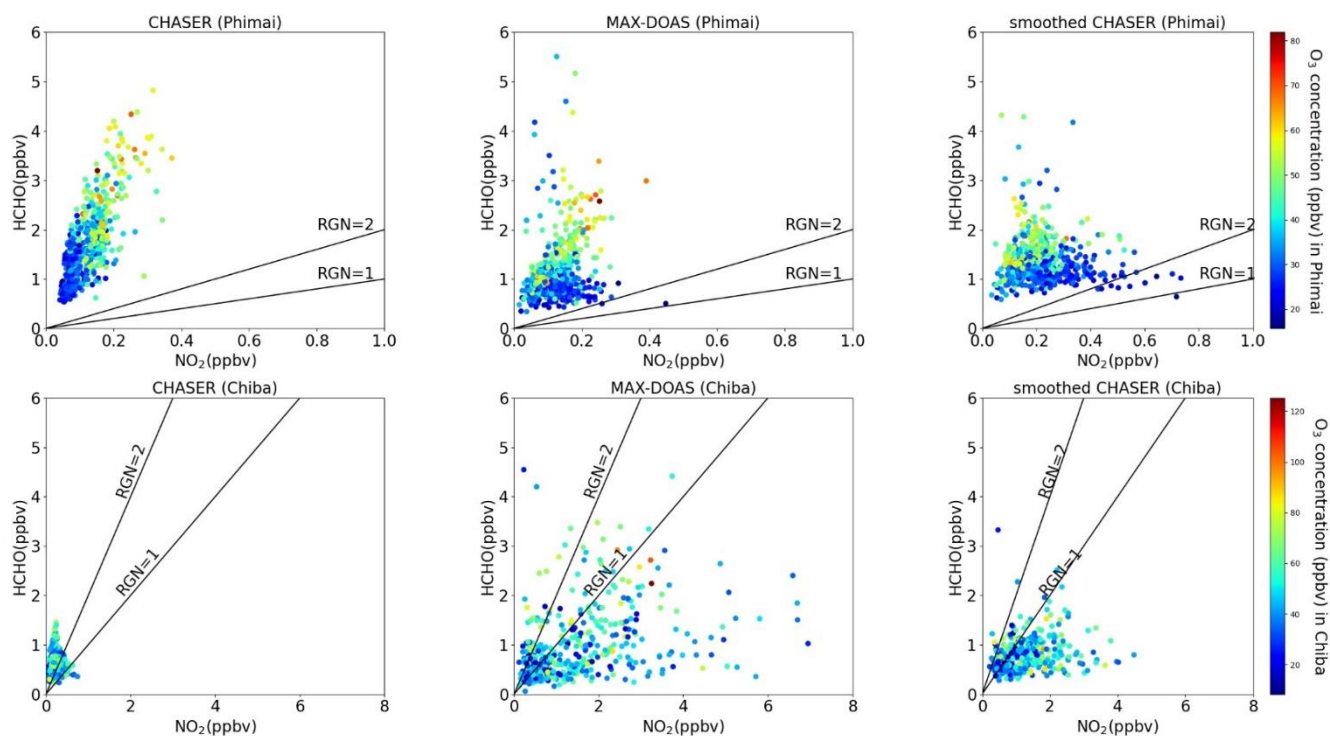
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3.2.4 Comparison between CHASER and MAX-DOAS R_{FN} values:

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Scatter plots of the CHASER and MAX-DOAS HCHO and NO₂ abundances in the 0–3 km layer of the troposphere color-coded with MAX-DOAS O₃ concentrations for the Phimai and Pantnagar site are depicted in Fig. 14. This discussion excludes the Pantnagar site because of challenges that hinder application of the AK, as discussed in section 3.2.1. The O₃ production regime in Phimai is NO_x limited. Also, CHASER produced similar results for the Phimai site, with high O₃ concentrations associated primarily with NO₂ concentrations < 0.4 ppbv. Some discrepancies were found between CHASER and MAX-DOAS NO₂ abundances during the wet season, leading to differences in the absolute R_{FN} values. Such discrepancies in R_{FN} values are minimized by smoothing the CHASER output with that of the MAX-DOAS AK.



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Figure 14. The scatter plots of the daily mean HCHO and NO₂ concentrations in the 0–3 km layer at Phimai and Chiba, inferred from MAX-DOAS observations, CHASER simulations, and smoothed CHASER output (with MAX-DOAS AK), coloured with the daily mean O₃ concentrations in the 0–3 km layer retrieved from the MAX-



DOAS observations at the respective sites. The solid lines indicate the R_{GN} values at 2 and 1, respectively. In addition, the time and date of the CHASER simulation, which coincides with the MAX-DOAS observations at the respective sites were selected.

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At the Chiba site, CHASER and MAX-DOAS R_{FN} values differ considerably, indicating the O_3 production regime. The smoothed CHASER output shows that the R_{FN} values are mostly lower than 1, which is indicative of a VOC limited region, consistent with the observations and the results reported by Irie et al. (2021). At the Chiba site, there are days (~ 80) when the O_3 sensitivity lies in the transition region (i.e., $1 < R_{FN} < 2$). However, the R_{FN} values were less than unity most days (~ 280), suggesting a dominant VOC limited region.

807

808

809 Conclusions

MAX-DOAS observations were made at three A-SKY sites from January 2017 through December 2018. Also, NO_2 and HCHO concentrations and profiles were retrieved using the JM2 algorithm. The retrieved products were compared with those of simulations obtained using the global chemistry transport model CHASER. At all three sites, the seasonal variation of both trace gases was consistent throughout the investigated period. At Phimai and Pantnagar, biomass burning led to enhanced HCHO and NO_2 concentrations, respectively, during the dry season and spring and post-monsoon season. Both the model and observations confirmed the contribution of soil NO_x emissions to the NO_2 concentrations in Phimai. The R_{FN} ratio estimated from the observations showed that the O_3 production regime in Phimai and Pantnagar was NO_x limited, whereas the O_3 production in Chiba was governed primarily by the NO_x emissions (VOC limited). The observed and modeled R_{FN} ratios estimated for the Phimai and Chiba sites



were consistent. CHASER demonstrated good capability for reproducing the HCHO seasonal variation at Phimai and Chiba. Although the HCHO datasets for the Pantnagar site showed discrepancies, the modeled HCHO seasonal variation in the IGP region was consistent with values found in the literature, indicating that the differences were related mainly to the model's coarse resolution.

Moreover, general overestimation was made of the biogenic emissions in the model. The observed and modeled NO₂ variation in Phimai and Chiba showed good agreement. However, the a priori values had a marked effect on the smoothed model output. In Phimai, the biomass burning contributions to the HCHO and NO₂ concentrations were up to ~50 and ~35%, respectively, whereas, in Pantnagar, the pyrogenic contributions were ~13 and ~8%, respectively. The estimates for Pantnagar were underestimated, as inferred from observationally obtained evidence, which can be improved with model simulations at higher spatial resolution. Anthropogenic emissions were revealed as an important driver of the VOC variation in the IGP region after biogenic emissions, which is consistent with findings reported in the literature.

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Code availability. The CHASER and JM2 source codes are not available publicly. Dr. Kengo Sudo (kengo@nagoya-u.jp) is the contact person for readers and researchers interested in the CHASER model. In addition, Dr. Hitoshi Irie (hitoshi.irie@chiba-u.jp) will answer queries regarding the usage of the JM2 codes.

839

Data availability: The MAX-DOAS data used in the study are publicly accessible on the A-SKY network website (<http://atmos3.cr.chiba-u.jp/a-sky/data.html>). The corresponding author can provide the CHASER simulations and MAX-DOAS averaging kernel data upon request.

843



844 *Author contributions:* HMSH conceptualized the study, conducted the model simulations, analyzed the
845 data, and drafted the manuscript. AMF helped with the data processing. HI developed the JM2 code and
846 maintained the A-SKY network. KS developed the CHASER model and supervised the study. AD shared
847 his experience to explain the results. HI, KS, AD, and AMF commented and provided feedback on the
848 final results and manuscript.

849

850 *Conflict of Interest:* The authors declare that they have no conflict of Interest

851

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856 Tsukuba, Japan.

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