



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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17 **Abstract.** Formaldehyde (HCHO) and nitrogen dioxide (NO₂) concentrations and profiles were retrieved
18 from ground-based multi-axis differential optical absorption spectroscopy (MAX-DOAS) observations
19 during January 2017 through December 2018 at three sites in Asia: (1) Phimai (15.18°N, 102.5°E),
20 Thailand; (2) Pantnagar (29°N, 78.90°E) in the Indo Gangetic plain (IGP), India; and (3) Chiba (35.62°N,
21 140.10°E), Japan. The observations were used to evaluate the NO₂ and HCHO partial columns and
22 profiles (0-4 km) simulated using the global chemistry transport model (CTM) CHASER. The NO₂ and
23 HCHO concentrations at all three sites showed consistent seasonal variations throughout the investigated
24 period. Biomass burning affected the HCHO and NO₂ variation in Phimai during the dry season and in
25 Pantnagar during spring (March-May) and post-monsoon (September-November). The results on the
26 HCHO to NO₂ ratio (R_{FN}), an indicator of high ozone sensitivity, show that the transition region (i.e., $1 <$
27 $R_{FN} < 2$) changes regionally, echoing the recent finding on the effectiveness of R_{FN} . Moreover, reasonable
28 estimates of transition regions can be derived accounting for the NO₂- HCHO chemical feedback.



29 CHASER demonstrated good performances reproducing the HCHO and NO₂ abundances at Phimai,
30 mainly above 500 m from the surface. Model results agree with the measured variations, ranging within
31 the one sigma standard deviation of the observations. Despite the complex terrain of Pantnagar
32 (mountainous terrain), the modeled NO₂ estimates between 1.8 – 2km were reasonable. Simulations at
33 higher resolution improved the modeled NO₂ estimates in Chiba, reducing the mean bias error (MBE) in
34 the 0-2 km height by 35%. However, resolution-based improvements were limited to the surface layers.
35 Sensitivity studies showed pyrogenic emissions in Phimai contribute to the HCHO and NO₂
36 concentrations up to ~ 50 and ~35%, respectively.

37 **1 Introduction**

38 Formaldehyde (HCHO) is the most abundant carbonyl compound in the atmosphere. It is a high-yield
39 product of oxidization of all primary volatile organic compounds (VOCs) emitted from natural and
40 anthropogenic sources by hydroxyl radicals (OH). Oxidation of long-lived VOCs such as methane
41 produces a global HCHO background concentration of 0.2–1.0 ppbv in remote marine environments
42 (Weller et al., 2000; Burkert et al., 2001; Singh et al., 2004; Sinreich et al., 2005). Aside from oxidation
43 of VOCs, the significant sources of HCHO are direct emission from biomass burning, industrial processes,
44 fossil fuel combustion (Lee et al., 1997; Hak et al., 2005; Fu et al., 2008;), and vegetation (Seco et al.,
45 2007). However, oxidization of non-methane VOCs emitted from biogenic (e.g., isoprene) or
46 anthropogenic (e.g., butene) sources govern the spatial variability of HCHO on a global scale (Franco et
47 al., 2015). The sinks of HCHO include photolysis at wavelengths below 400 nm, oxidation by OH, and
48 wet deposition, thereby limiting the lifetime to a few hours (Arlander et al., 1995).

49 Nitrogen dioxide (NO₂) is an important atmospheric constituent that (1) participates in the catalytic
50 formation of tropospheric ozone (O₃), (2) acts as a catalyst for stratospheric ozone (O₃) destruction
51 (Crutzen, 1970), (3) contributes to the formation of aerosols (Jang and Kamens, 2001), (4) acts as a
52 precursor of acid rain (Seinfeld and Pandis, 1998), and (5) strongly affects radiative forcing (Lelievre et
53 al., 2002; Solomon et al. 1999). Nitrogen oxides (NO_x = NO (nitric oxide) + NO₂) are emitted from natural
54 and anthropogenic sources. Primary NO_x emission sources are biomass burning, fossil fuel combustion,
55 soil emissions, and lightning (Bond et al., 2001; Zhang et al., 2003). NO_x emissions degrade air quality
56 and are a leading air pollutant with global and national ambient air quality standards (Ma et al., 2013).



57 Both HCHO and NO₂ are important intermediates in the global VOC–HO_x (hydrogen oxides)–NO_x
58 catalytic cycle, which governs the O₃ chemistry in the troposphere (Lee et al., 1997; Houweling et al.,
59 1998; Kanakidou et al., 2005; Hak et al., 2005). Thus, both trace gases play crucial roles in tropospheric
60 chemistry.

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

71 The ability of MAX-DOAS to provide information related to surface concentrations, vertical profiles,
72 and column densities makes it complementary to ground-based in situ and satellite observations.
73 Moreover, the MAX-DOAS method uses narrowband absorption of the target compounds, thereby
74 obviating the radiometric calibration of the instrument. Because of these advantages, MAX-DOAS
75 systems are deployed for the assessment of aerosol and trace gases in regional and global observational
76 networks such as BREDOM (Wittrock et al., 2004), BIRA-IASB (Clémer et al., 2010), and MADRAS
77 (Kanaya et al., 2014). Such datasets are used in but are not limited to (1) air quality assessment and
78 monitoring, (2) evaluation of chemistry-transport models (CTMS), and (3) validation of satellite retrieval.
79 Several studies have used MAX-DOAS datasets to validate tropospheric columns retrieved from satellite
80 observations, including NO₂ and HCHO (Ma et al., 2013; Irie et al., 2008b; Chan et al., 2020; Ryan et al.,
81 2020). However, limited MAX-DOAS datasets have been used to evaluate global CTMs. Vigouroux et
82 al. (2009) and Franco et al. (2015) respectively used the MAX-DOAS HCHO datasets from Reunion
83 Island and Jungfraujoch stations to evaluate the IMAGES (Intermediate Model of Annual and Global
84 Evolution of Species) and GEOS-Chem model simulations. Kanaya et al. (2014) validated the MIROC-



85 ESM-CHEM (Model for Interdisciplinary Research on Climate–Earth System Model – Chemistry)
86 simulated NO₂ column densities with MAX-DOAS observations in Cape Hedo and Fukue in Japan.
87 Kumar et al. (2021) utilized MAX-DOAS observations to evaluate the high-resolution regional model
88 MecO(n) (MESSy-field ECHAM and COSMO model nested n times).
89 In this study, NO₂ and HCHO profiles retrieved from MAX-DOAS observations at the A-SKY
90 (International air quality and sky research remote sensing) (<http://atmos3.cr.chiba-u.jp/a-sky/>) network
91 sites are utilized to evaluate the global Chemical Atmospheric General Circulation Model for the Study
92 of Atmospheric Environment and Radiative Forcing (CTM CHASER; Sudo et al., 2002). The three A-
93 SKY sites - (1) Phimai in Thailand (15.18°N, 102.56°E), (2) Pantnagar (29°N, 78.90°E) in the Indo
94 Gangetic plain (IGP) in India, and (3) Chiba (35.62°N, 140.10°E) in Japan, are representative of rural,
95 semi-rural, and urban environments, respectively. CHASER has been used mostly for global-scale
96 research (Sudo et al., 2007; Sekiya et al., 2014, 2018; Miyazaki et al., 2017). This report is the first attempt
97 to evaluate the CHASER-simulated NO₂ and HCHO profiles using MAX-DOAS observations in three
98 atmospheric environments. Moreover, limited literature has described the use of MAX-DOAS datasets to
99 evaluate global CTMs in the south and southeast Asian regions. Overall, this study was conducted to
100 provide important insights into the model performances and help reduce model uncertainties related to
101 NO₂ and HCHO simulations in these regions.

102 The manuscript is structured in the following manner. First, the observation sites, MAX-DOAS
103 instrumentation, and retrieval strategies are described in section 2. Section 2 also includes a short
104 description of the CHASER model. Next, the observations and the evaluation of the model are presented
105 in sections 3.1 and 3.2, respectively. Finally, the sensitivity study results are provided in section 3.3. and
106 the concluding remarks in section 4.

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108



109 **2 Observations, datasets, and methods**

110

111 **2.1 Site Information**

112 Continuous MAX-DOAS observations at Phimai, Pantnagar, and Chiba started in 2014, 2017, and 2012,
113 respectively. The measurements from January 2017 to December 2018 at all three sites are discussed here.
114 Phimai, a rural site, is located ~260 km northeast of the Bangkok metropolitan region and is unlikely to
115 be affected by vehicular and industrial emissions. However, the site is affected by biomass burning from
116 January to April. Two major air streams: the dry, cool northeast monsoon during November – mid-
117 February and the wet, warm southwest monsoon during mid-May – September affect the climate in
118 Phimai. Following Hoque et al. (2018), the climate classification of Phimai is : (a) dry season (January –
119 April), and (b) wet season (June – September).

120 Pantnagar, a semi-urban site in India, is located in the Indo-Gangetic Plain region (IGP). The Indian
121 capital of New Delhi is situated at ~225 km southwest of the site. The low-altitude plains are on the south
122 and west side of the site, whereas the Himalayan mountains are located to the north and east. An important
123 roadway with moderate traffic volume and a small local airport lies within 3 km of the site. Rudrapur
124 (~12 km southwest of Pantnagar) and Haldwani (~ 25 km northeast of Pantnagar) are the two major cities
125 near Pantnagar, having a few small-scale industries. Climate classification in Pantnagar is the following:
126 (1) winter (December–February), (2) spring (March–May), (3) summer monsoon (June–August), and (4)
127 autumn (September–November).

128 Chiba, an urban site, is located ~40 km southeast of the Tokyo metropolitan region. Tokyo Bay, large-
129 scale industries, and residential areas are located within a 50 km radius. The climate in Chiba is divided
130 into four seasons: (1) spring (March–May), (2) summer (June–August), (3) autumn (September–
131 November), and winter (December– February). The locations of the three sites are depicted in Fig. 1.

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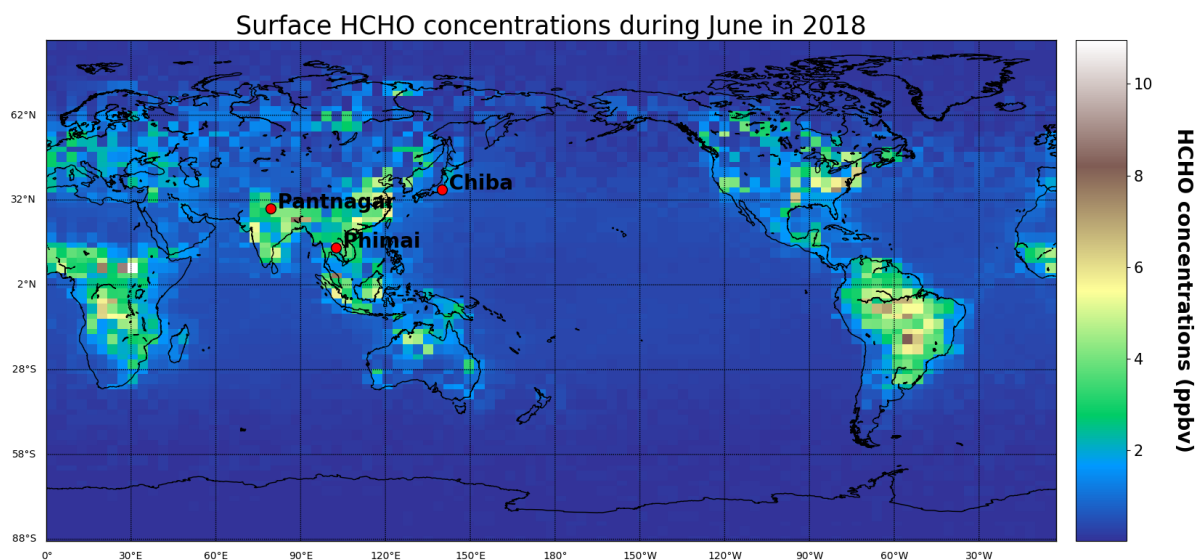
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137 **Figure 1:** Surface HCHO concentrations during June 2018, simulated using the CHASER model. The
138 red points indicate the locations of the observation sites, which are part of the A-SKY network.

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141 **2.2 MAX-DOAS retrieval**

142 The MAX-DOAS systems used for continuous observations at the three sites participated in the Cabauw
143 Intercomparison Campaign of Nitrogen Dioxide measuring Instruments (CINDI) (Roscoe et al., 2010)
144 and CINDI-2 (Kreher et al., 2020) campaign. The instrumentation setup is described by Irie et al. (2008,
145 2011, 2015). The indoor part of the MAX-DOAS systems consists of an ultraviolet-visible (UV-VIS)
146 spectrometer (Maya2000Pro; Ocean Optics Inc.) embedded in a temperature-controlled box. The outdoor
147 unit consist of a single telescope and a 45° inclined movable mirror on a rotary actuator, used to perform
148 reference and off-axis measurements. The high-resolution spectra from 310–515 nm is recorded at six



149 elevation angles (ELs) of 2°, 3°, 4°, 6°, 8°, and 70° at the Chiba and Phimai sites. At the Pantnagar site,
150 measurements are conducted at ELs of 3°, 4°, 5°, 6°, 8°, and 70°. The sequences of the ELs at all the sites
151 were repeated every 15 min. The reference spectra is recorded at EL of 70° instead of 90° to minimize
152 variations in the measured signals. The off-axis ELs is limited to < 10° to reduce the systematic error in
153 the in-oxygen collision complex (O₄) fitting results (Irie et al., 2015), thereby maintaining high sensitivity
154 in the lowest layer of the retrieved aerosol and trace gas profiles. Daily wavelength calibration using the
155 high-resolution solar spectrum from Kurucz et al. (1984) is performed to account for the spectrometer's
156 long-term degradation. The spectral resolution (full width half maximum: FWHM) is about 0.4 nm at 357
157 and 476 nm. The concentrations and profiles of aerosol and trace gases are retrieved using the Japanese
158 vertical profile retrieval algorithm (JM2 ver. 2) (Irie et al., 2011, 2015). The algorithm works in three
159 steps: (1) DOAS fittings, (2) profile/column retrieval of aerosol, and (3) profile/column retrieval of trace
160 gases. Irie et al. (2008a, 2008b, 2011, 2015) described the retrieval procedures, and the error estimates.
161 Here we provide a short overview.

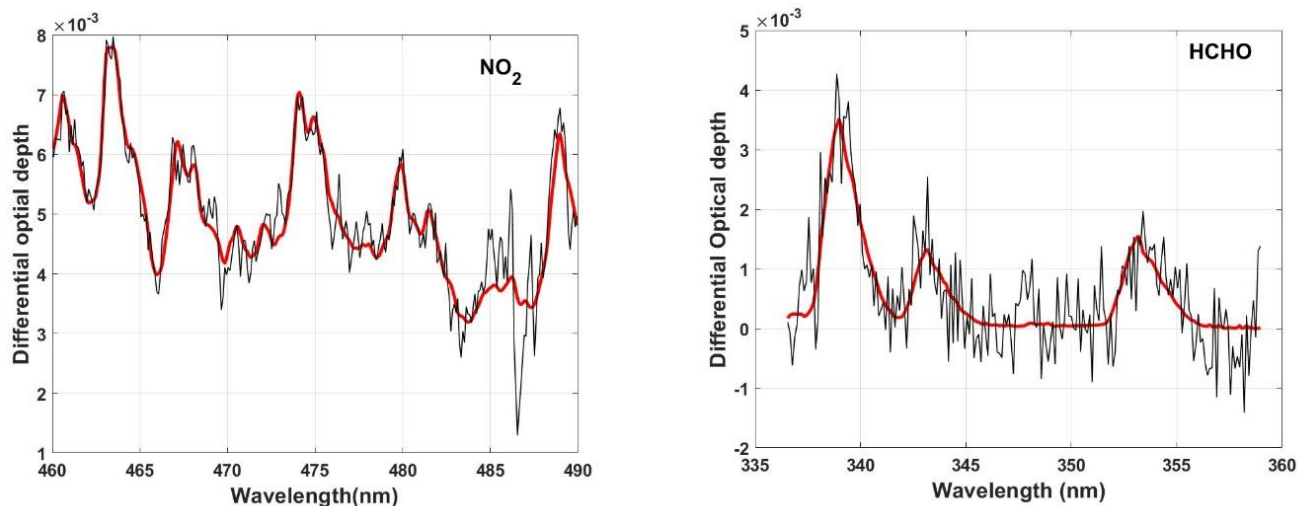
162 First, the differential slant column density (ΔSCD) of trace gases is retrieved using the DOAS
163 technique (Platt 1994), which uses the nonlinear least-squares spectral fitting method, according to the
164 following equation.

$$165 \quad \ln I(\lambda) = \ln(I_o(\lambda) - c(\lambda)) - \sum_i^n \sigma_i(\lambda) \Delta\text{SCD}_i - p(\lambda) \quad (1)$$

166
167 Therein, $I_o(\lambda)$ represents the reference spectrum measured at time t . $I_o(\lambda)$ is derived by interpolating
168 two reference spectra (i.e., EL=70°) within 15 min before and after the complete sequential scan of the
169 off-axis ELs at time t . ΔSCD represents the difference between the slant column density along the off-
170 axis and reference spectrum. Second- and third-order polynomials are fitted to account for the
171 wavelength-dependent offset $c(\lambda)$ and the effect of molecular and particle scattering $p(\lambda)$, respectively.
172 Also, $c(\lambda)$ accounts for the influence of stray light. The HCHO ΔSCD and NO₂ ΔSCD are retrieved from
173 the fitting windows of 336–359 and 460–490 nm, respectively. Significant O₄ absorptions in the 338–
174 370 and 460–490 nm fitting windows are utilized to retrieve the O₄ ΔSCDs . The absorption cross-section



175 data and the fitted absorbers in the HCHO and NO₂ fitting windows are given in table 1. Figure 2 presents
 176 an example of the fitting results.



177 **Figure 2:** Examples of spectral fitting of NO₂ and HCHO, where the red and black lines show the scaled cross-
 178 section and the summation of scaled cross-sections and fitting residuals, respectively. The example is shown for
 179 the measurements on 10 April 2017, in Phimai at 10:00 LT at an EL of 2°.

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

184

185
$$y = (O_4 \Delta SCD(\Omega_1) \dots \dots \Delta SCD(\Omega_n))^T \quad (2) \text{ and}$$

186
$$x = (\tau F_1 F_2 F_3)^T \quad (3),$$



187

188 **Table 1.** Cross section data references and absorbers fitted in the HCHO and NO₂ window

Cross-section	Absorbers fitted	Data Source
O ₃		<i>Bougmil et al.</i> [2003], 223K
NO ₂	O ₃ , NO ₂ , H ₂ O, O ₄ , Ring	<i>Vandaele et al.</i> [1996], 295K
BrO		<i>Fleischmann et al.</i> [2004], 223K
Ring		<i>Chance and Spurr</i> [1997]
H ₂ O		<i>Vandaele et al.</i> [2005], 280K
O ₄		Hermans et al. [2003], 296K
HCHO	O ₃ , NO ₂ , HCHO, BrO, O ₄ , Ring	<i>Meller and Moortgart</i> [2000], 293k

189

190

191 where n stands for the number of measurements within one complete scan of an EL sequence. Ω denotes
192 the viewing geometry and compromise three components: solar zenith angle (SZA), EL, and relative
193 azimuth angle (RAA). The F values determine the profile shape, with values between 0 and 1. The partial
194 AOD for 0–1, 1–2, 2–3, and above 3 km layers were defined respectively as $\text{AOD} \cdot F_1$, $\text{AOD} \cdot (1-F_1) F_2$,
195 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 3 to 100 km is derived
196 assuming a fixed value at 100 km and exponential AEC profile shape. Similarly, the AEC profiles at 2–
197 3, 1–2, and 0–1 km were derived. Such parameterization provides the advantage that the AEC profile can
198 be retrieved with only the apriori knowledge of the F (profile shape) values and little or no information
199 related to the absolute AEC values in the troposphere. Irie et al. (2008a) demonstrated that the relative



200 variability of the profile shape, in terms of 1-km averages, is smaller than that of the absolute AEC values.
201 However, the vertical resolution and the measurement sensitivity cannot be derived directly with such a
202 parameterization (Irie et al., 2008a; 2009). The retrievals and simulations conducted by other groups for
203 similar geometries (i.e., Frieß et al., 2006) are used to overcome such limitations. The apriori values used
204 for this study were similar to those reported by Irie et al. (2011): $AOD = 0.21 \pm 3.0$, $F_1 = 0.60 \pm 0.05$, F_2
205 $= 0.80 \pm 0.03$, and $F_3 = 0.80 \pm 0.03$.

206 Then, a lookup table (LUT) of the box air mass factor (A_{box}) vertical profile is constructed using the
207 radiative transfer model JACOSPAR (Irie et al., 2015), which is based on the Monte Carlo Atmospheric
208 Radiative Transfer Simulator (MCARaTS) (Iwabuchi, 2006). Results obtained from JACOSPAR are
209 validated in the study of Wagner et al. (2007). The optimal aerosol load and the A_{box} profiles are derived
210 using the A_{box} LUT and the $O_4 \Delta SCD$ at all ELs.

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

214

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

$$216 \quad y = (NO_2 \Delta SCD(\Omega_1) \dots \dots \dots NO_2 \Delta SCD(\Omega_n))^T \quad (4) \text{ and}$$

217

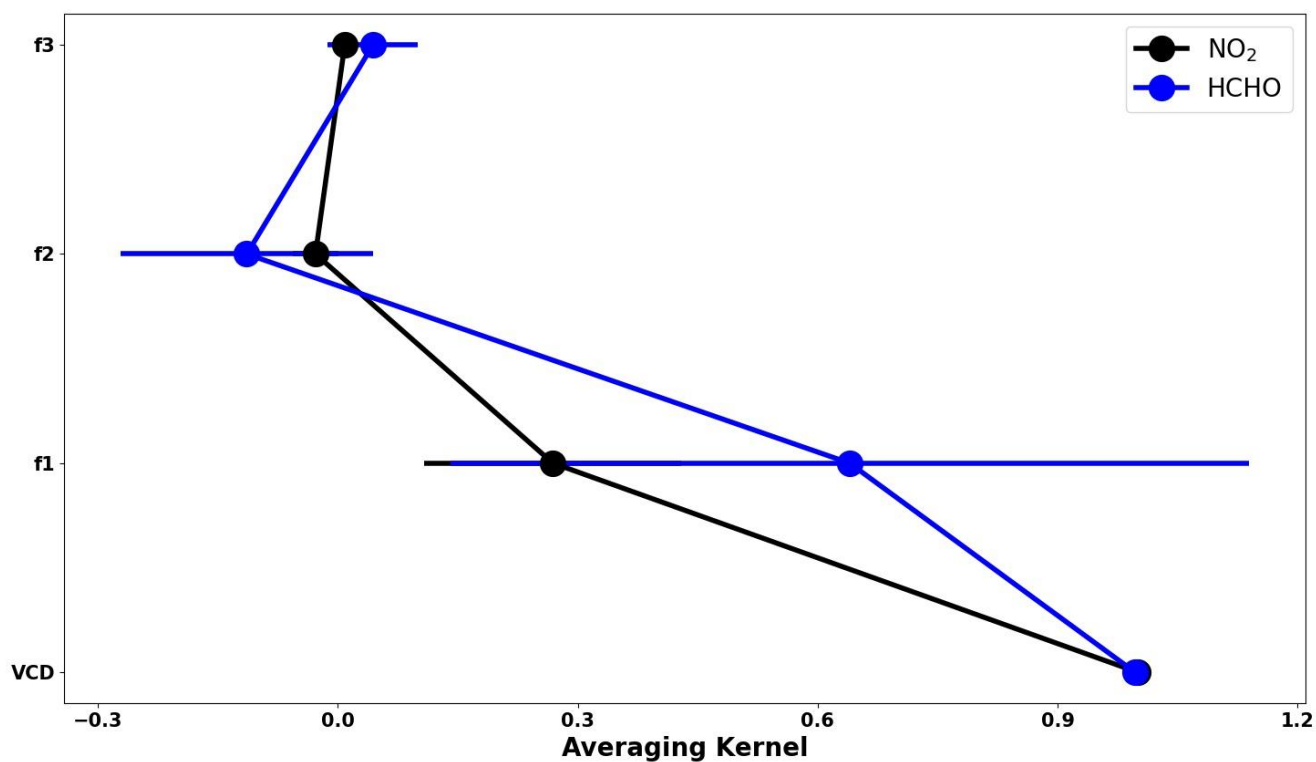
$$218 \quad x = (VCD f_1 f_2 f_3)^T \quad (5)$$

219 VCD represents the vertical column density below 5 km; f values are the profile shape factor. Above
220 the 5 km layer, fixed profiles are assumed. Similarly, to aerosol retrieval, the partial VCD for the 0–1, 1–
221 2, 2–3, and 3–5 km is 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
222 $\cdot (1-f_1) (1-f_2) (1-f_3)$. Finally, the partial VCD values are converted to the volume mixing ratio (VMR) using
223 the U.S. standard atmosphere temperature and pressure data scaled to the respective surface
224 measurements.

225 The calculated vertical profile is converted to $NO_2 \Delta SCDs$ using the A_{box} LUT constructed for aerosol
226 retrieval. However, the trace gas wavelengths differed from the representative wavelengths of A_{box} LUT
227 (357 and 476 nm). Therefore, the AOD at the trace gas wavelength is estimated, converting the retrieved



228 AOD to the closer aerosol wavelength of 357 or 476 nm, assuming the Angstrom exponent value of 1.00.
229 Then, the A_{box} profiles from the LUT corresponding to the recalculated AOD values are selected. The
230 dependence of the A_{box} profiles on the concentration profiles is expected to be minimal because both
231 HCHO and NO_2 are optically thin absorbers (Wagner et al., 2007; Irie et al., 2011). For every 15 min
232 (time required for one complete scan of ELs), 20% (the mean ratio of the retrieved VCD to maximum
233 ΔSCD) of the maximum trace gas ΔSCD s is used as a priori information for the VCD retrievals. The a
234 priori error is set to 100% of the maximum trace gas ΔSCD . Figure 3 presents the mean averaging kernel
235 of the HCHO and NO_2 retrievals during the dry season in Phimai. The area calculated from the averaging
236 kernel is close to unity. Therefore, the retrieved VCD is independent of the a priori values.
237



238
239 **Figure 3:** The mean averaging kernel of the NO_2 and HCHO retrievals from the observations in Phimai during
240 2017
241



242 The total error of the retrieval consists of random and systematic errors. The measurement error
243 covariance matrix constructed from the residuals of the respective trace gas Δ SCDs is used to estimate
244 the random error. The systematic error is calculated assuming uncertainties as high as 30 and 50% in the
245 retrieved AOD (or the corresponding A_{box} values). The total estimated error is shown in Table 2. Aside
246 from the random and systematic error, more sources of error might exist. For instance, the bias in the ELs
247 can induce uncertainties in the retrieved products. However, Hoque et al. (2018) demonstrated that such
248 biases had a non-significant effect on the final retrieved products, mostly less than 5%.

249 The cloud screening procedure is similar to that described by Irie et al. (2011) and by Hoque et al.
250 (2018a, 2018b). During the retrieval steps, retrieved AOD values greater than 3 are excluded. This is
251 because optically thick clouds are primarily responsible for such large optical depth. Furthermore,
252 filtering based on the residuals of O_4 and the trace gas Δ SCDs is also used to screen clouds. The screening
253 criteria are: respective residuals of O_4 , HCHO, and NO_2 Δ SCDs < 10%, < 50%, and < 20%, and the degrees
254 of freedom of the retrieval greater than 1.02.

255

256 **Table 2.** Estimated Errors (%) for the NO_2 and HCHO concentration in 0-1 km layer, retrieved using the
257 JM2 algorithm

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

258

259

260 2.3 CHASER simulations

261 CHASER V4.0 (Sudo et al., 2002; Sudo and Akimoto, 2007; Sekiya and Sudo, 2014), coupled online
262 with the MIROC-AGCM atmospheric general circulation model (AGCM) (K-1 model developers, 2004)
263 and the SPRINTARS aerosol transport model (Takemura et al., 2005, 2009), is a global chemistry
264 transport model to study the atmospheric environment and radiative forcing. In addition, several updates,



265 including the introduction of aerosol species (sulfate, nitrate, etc.) and related chemistry, radiation, and
 266 cloud processes, have been implemented in the latest version of CHASER.

267 CHASER can calculate the concentrations of 92 species through 263 chemical reactions (gaseous,
 268 aqueous, and heterogeneous chemical reactions) considering the chemical cycle of O_3 – HO_x – NO_x – CH_4 –
 269 CO along with oxidation of non-methane volatile organic compounds (NMVOCs)(Miyazaki et al., 2017).
 270 CHASER simulates the stratospheric ozone chemistry considering the Chapman mechanisms, catalytic
 271 reactions related to halogen oxides (HO_x , NO_x , ClO_x , and BrO_x), and polar stratospheric clouds (PSCs).
 272 Resistance-based parameterization (Wesely, 1989), cumulus convection, and large-scale condensation
 273 parameterizations are used to calculate dry and wet depositions. The piecewise parabolic method (Colella

274 **Table 3:** The settings of the CHASER simulations used in the current study

Simulation	Anthropogenic emissions	Pyrogenic emissions	Biogenic emissions	Soil NO_x emission	Other physical and chemical processes
Standard	ON	ON	ON	ON	ON
L1_HCHO	ON	Pyrogenic VOCs switched	ON	ON	ON
L1_opt	ON	OFF	Reduced by 50%	ON	ON
L1_NO₂	ON	ON	ON	OFF	ON
L2		ON	ON	ON	ON



Anthropogenic
VOC emissions
switched OFF

275

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

283 The anthropogenic, biomass burning, lightning, and soil emissions of NO_x were incorporated into
284 CHASER simulations. Anthropogenic emissions were based on the HTAP_v2.2 for 2008. Biomass
285 burning and soils emissions from the ECMWF/MAC reanalysis were used. The biogenic emissions for
286 VOCs are based on the process-based biogeochemical model the Vegetation Integrative Simulator for
287 Trace gases (VISIT) (Ito and Inatomi, 2012) simulations. The NO_x production from lightning is calculated
288 based on the parameterization of Price and Rind (1992) linked to the convection scheme of the AGCM
289 (Sudo et al., 2002). The isoprene, terpene, acetone, and ONMV emissions during July were 2.14×10^{-11} ,
290 4.43×10^{-12} , 1.60×10^{-12} , and $9.93 \times 10^{-13} \text{ kgCm}^{-2}\text{s}^{-1}$.

291 Multiple CHASER simulations with different settings used for this study are presented in Table 3.

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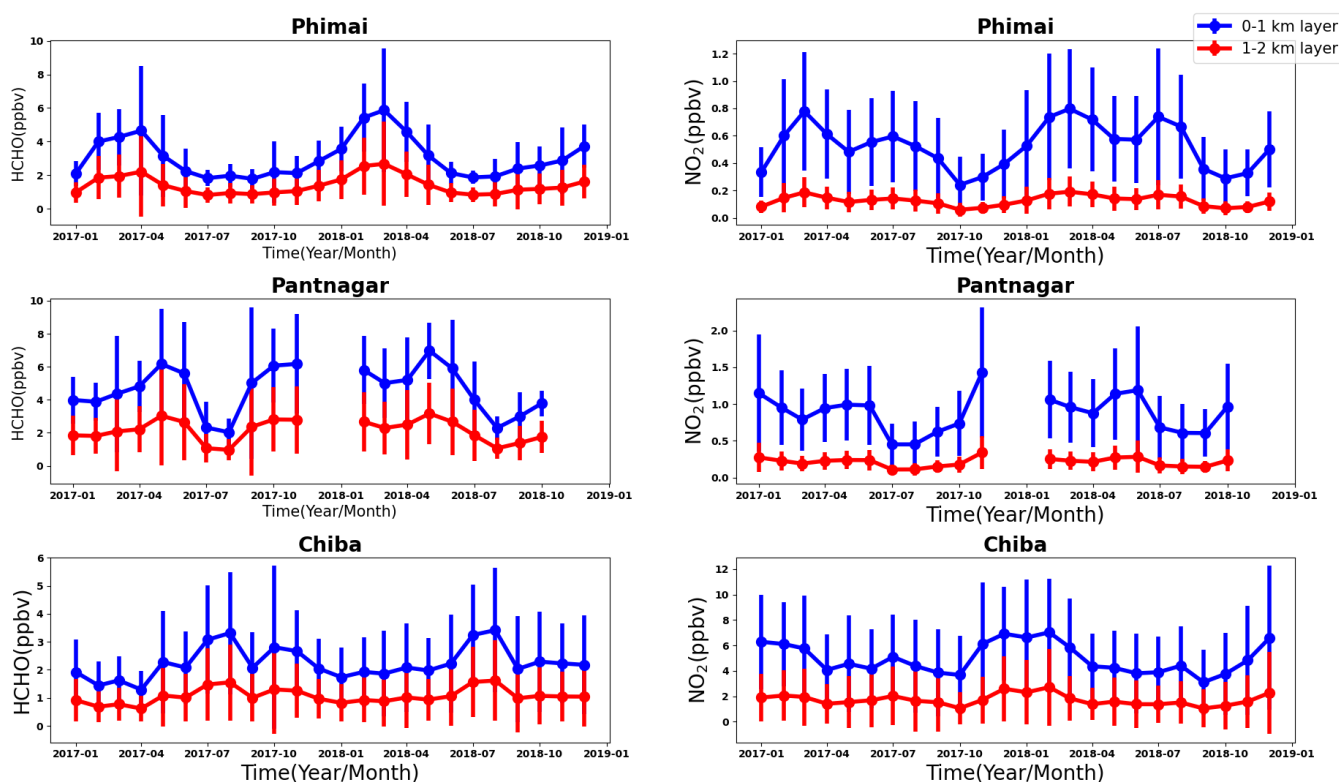
303 **3 Results and discussion**

304 **3.1 Results from MAX-DOAS observations**

305 **3.1.1 HCHO seasonal variation**



306 The monthly mean HCHO concentrations in the 0–1 and 0–2 km layers from January 2017 – December
307 2018 and the corresponding one sigma (1σ) standard deviations for the three sites are depicted in Fig. 4.
308 The HCHO levels at the Phimai site show a consistent seasonal cycle, characterized by high
309 concentrations during the dry season. Such enhancement is related to the influence of biomass burning
310 during the dry season, which has been well documented in the work of Hoque et al. (2018). The HCHO
311 concentrations at Phimai peak in March or April, with a maximum of 4–6 ppbv. The variation in the peak
312 concentration and timing mainly depends on the intensity of biomass burning activities. During the wet
313 season, the HCHO concentrations are mostly within 2–3 ppbv, indicating a two-fold increase in the
314 HCHO abundances during the dry season. The daily mean HCHO amounts (0–1 km) are 0.78–9.84 ppbv,
315 representing seasonal modulation of 134%.



316
317 **Figure 4:** The seasonal variations in the HCHO (left panel) and NO₂ (right panel) concentrations in the 0–1 (blue)
318 and 1–2 (red) km layers in Phimai, Pantnagar, and Chiba. The error bars indicate the one sigma standard deviation



319 of the mean values. The gaps in the plots for the Pantnagar site indicate the unavailability of observations during
320 the investigated period.

321

322

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

340 Summertime maximum and wintertime minimum characterize the seasonal variations of HCHO at
341 the Chiba site, with a peak at ~3 ppbv. The HCHO concentrations are ~2 ppbv during other seasons,
342 similar to the HCHO concentrations in Phimai during the wet season. The seasonal variation amplitudes
343 of HCHO in Chiba is ~94%. For a site with similar seasonal variation (i.e., summertime maximum and
344 wintertime minimum), Franco et al. (2015) reported HCHO seasonal modulation of 88%.

345 The concentrations of HCHO in the 1–2 km layers at all three sites are lower, almost 50% the value
346 of the concentrations in the 0–1 km layer. The HCHO seasonal variation amplitudes at Phimai, Pantnagar,



347 and Chiba sites are, respectively, 131%, 102%, and 90% when calculated based on the HCHO
348 concentration in the 1–2 km layers. The modulation further decreased when retrieved values for the 2–3
349 km layer R used.

350 **3.1.2 NO₂ seasonal variation at the three sites**

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

359 At Phimai, the NO₂ concentrations in both seasons are almost similar. However, when Hoque et al.
360 (2018a) reported the seasonal variation of NO₂ at Phimai during 2015–2016, the NO₂ concentrations in
361 the dry season were higher. Table 4 shows the number of fire events during the dry seasons from 2015 to
362 2018. The fire data are extracted from the MODIS Active Fire Detections database
363 (<https://firms.modaps.eosdis.nasa.gov>, last accessed on 2021/12/15). Data fulfilling the following criteria
364 are chosen – (a) data point located within 100 km of the Phimai site, (b) confidence of the data greater
365 than 70%, and (c) observations during the daytime. The lower fire counts during 2017–2018 compared to
366 2015–2016 period coincide with the lower NO₂ concentrations in the former. Fire counts varied between
367 2017 and 2018 but did not impact the NO₂ levels. However, HCHO concentrations changed with the
368 number of fire occurrences between 2015 – 2018 (i.e., Figure 1 and Hoque et al., 2018a).

369 At such low NO₂ levels in Phimai, soil NO_x emissions are likely to be a more relevant contribution to
370 NO₂. Although NO₂ is not emitted directly from soils, biological processes emit NO, which rapidly
371 converts to NO₂ (Hall et al., 1996). In addition, many studies have established a relation between soil
372 moisture and NO emissions (Carden et al., 1993; Zheng et al., 2000; Schindlbacher et al., 2004; Huber et
373 al., 2020). The potential contribution of soil NO_x emissions, inferred from CHASER simulations, is
374 discussed in section 3.3.2.



375 **Table 4:** The number of fire events occurring during the dry season (January to April) in Phimai from
376 2015 to 2018. The selection criteria of the data are – (1) situated within 100 km of the site, (2) confidence
377 level > 70%, and (c) daytime measurements.

Dry season years	Number of fire events
2015	84
2016	98
2017	62
2018	77

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379

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381 3.1.3.1 The HCHO to NO₂ ratio (R_{FN}):

382 The HCHO to NO₂ (R_{FN}) ratio is considered as an indicator of high ozone O₃ sensitivity (Martin et al.,
383 2004; Duncan et al., 2010). The O₃ production regime is characterized as VOC-limited for $R_{FN} < 1$ and
384 NO_x- limited when $R_{FN} > 2$, and the values in the range 1-2 are said to be in the transition/ambiguous
385 region (Duncan et al., 2010; Ryan et al., 2020). After Tonnesen and Dennis (2000), several studies
386 utilized R_{FN} estimated from satellite and ground-based observations to infer O₃ sensitivity to NO_x and
387 VOCs (Irie et al., 2021; Jin and Holloway et al., 2015; Mahajan et al., 2015; Duncan et al., 2010; Martian
388 et al., 2004; etc.). However, the effectiveness of R_{FN} is still under discussion primarily based on two-
389 points- (1) the range of the transition region to categorize the VOC and NO_x -limited region, and (2) the
390 altitude dependence of R_{FN} (i.e., Jin et al. 2017). Most of the studies mentioned above used the transition
391 range ($1 < R_{FN} < 2$) proposed by Duncan et al. (2010). Schroeder et al. (2017) reported that a common
392 transition (i.e., $1 < R_{FN} < 2$) range might not be valid globally and should be calculated based on region.
393 At first, the results based on the standard transition range are discussed here, and then its applicability in
394 the study regions is inferred.

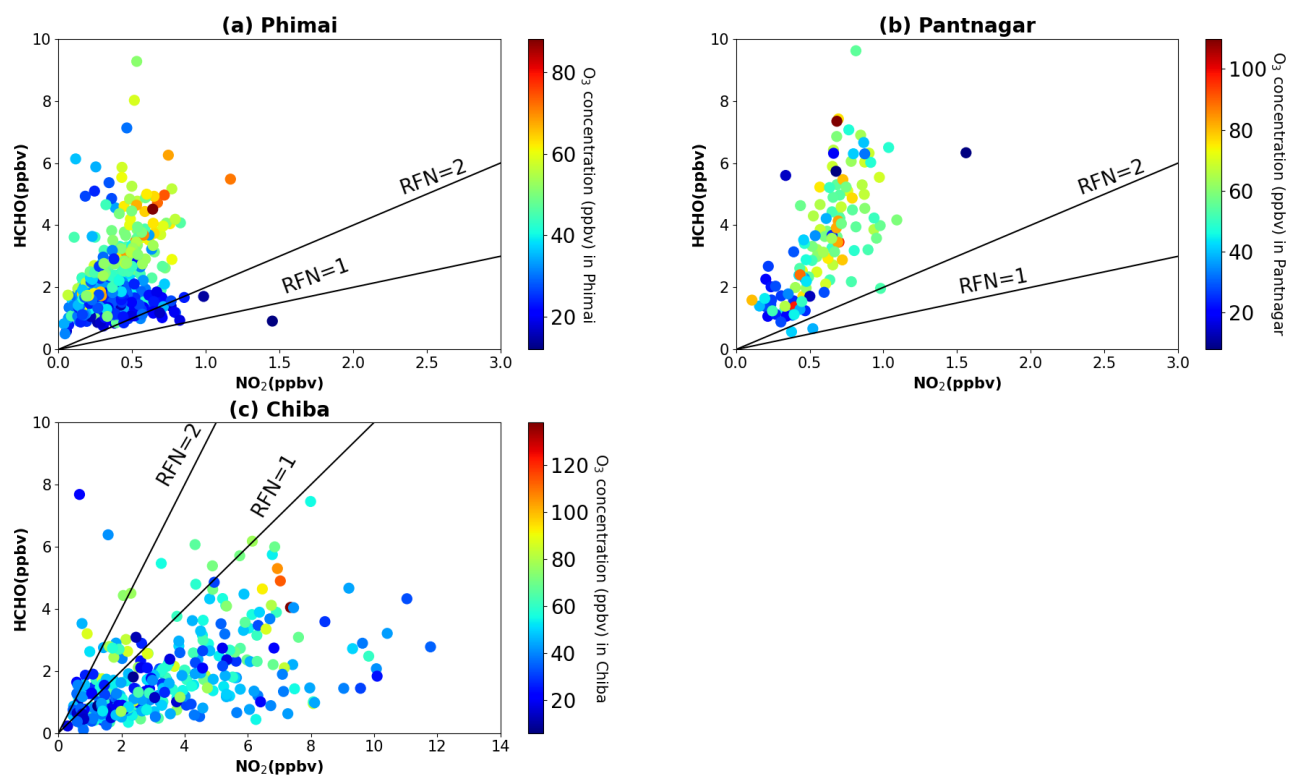
395 Figure 5 shows scatter plots of daily mean NO₂ and HCHO concentrations in the 0-2 km layer at the three
396 sites, color-coded with the respective O₃ concentrations (0-2 km). Retrieval of the JM2 O₃ product is
397 explained in Irie et al. (2011). The O₃ concentrations for SZA < 50° are used to minimize stratospheric



398 effects. The JM2 O₃ product showed good agreement with ozonesonde measurements (Irie et al., 2021).
399 The R_{FN} values for HCHO concentrations < 10 ppbv are shown, because the transition range $1 < R_{FN} < 2$
400 is valid for HCHO concentrations < 10 ppbv (Souri et al., 2020). Most of the high O₃ occurrences fall in
401 the $R_{FN} > 2$ region at Phimai and Pantnagar and in $R_{FN} < 1$ at Chiba. The common transition range
402 classifies the O₃ production regime as NO_x-limited in Phimai and Pantnagar and VOC-limited in Chiba.
403 The change of the upper limit of the HCHO concentrations does not significantly affect the results (Fig.
404 S1 in the supplementary information). At all sites, the R_{FN} values tend to be biased to a particular regime
405 (i.e., NO_x - or VOC-limited), with only 4 and 2% of the ratios in the range 0-2, in Phimai and Pantnagar,
406 respectively. This suggests that the transition occurs at higher or lower ratios than the common definition.
407 The recent work of Souri et al. (2020) found that the NO₂-HCHO relationship plays an important role in
408 determining the transition region and derived a formulation from accounting for the NO₂-HCHO chemical
409 feedback in the ratios as follows:

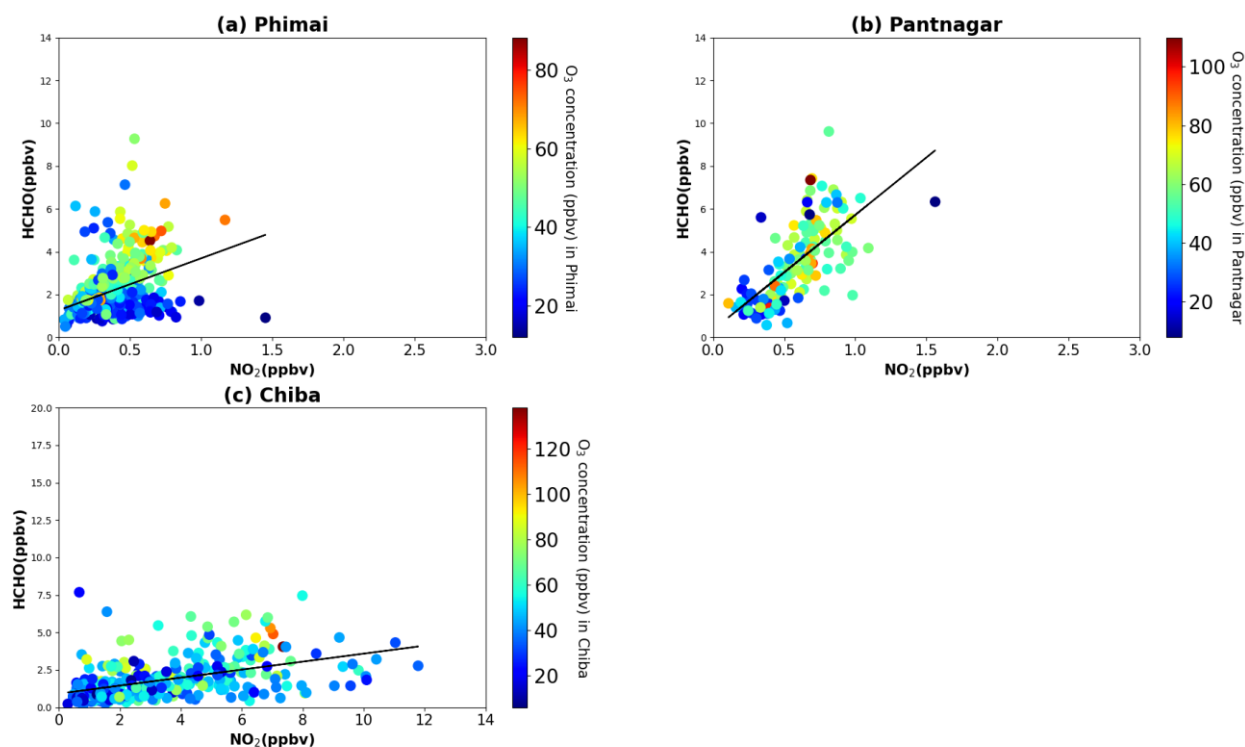
$$HCHO = m * (NO_2 - b) \quad (6)$$

410 where m and b are the slope and intercept. Equation (6) is based on observations, which means the
411 regionally adjusted fitting coefficients will reflect the local NO₂ - HCHO relationship. Solving equation
412 (6), the transition line estimated from the observations in the 0-2 km layer, is shown in Fig 6. Rather than
413 a range, the method calculates a single transition line, which corresponds to the NO₂-HCHO feedback.
414 The regions above and below the transition line are characterized as VOC- and NO_x -limited or other,
415 respectively.
416



417

418 **Figure 5.** The scatter plot of HCHO and NO₂ concentrations in the 0-2 km layer at (a) Phimai, (b) Pantnagar, and
419 (c) Chiba, colored with the O₃ concentrations in the 0-2 km layer at the respective sites. The solid lines indicate
420 $R_{FN} = 2$ and $R_{FN} = 1$ benchmark.



421

422 **Figure 6.** The scatter plot of HCHO and NO₂ concentrations in the 0-2 km layer at (a) Phimai, (b) Pantnagar, and
423 (c) Chiba, colored with the O₃ concentrations in the 0-2 km layer at the respective sites. The solid lines indicate the
424 transition lines calculated adopting the method of Sourì et al. (2020).

425

426 The revised transition line in Phimai and Pantnagar seems to be more reasonable than Fig. 5. In Phimai,
427 the transition line almost clearly distinguishes between the high and low O₃ occurrences. It is perceptible
428 that when the HCHO concentrations are higher than NO₂, the transition of the regimes is likely to occur
429 at higher R_{FN} values. The minimum and mean R_{FN} value along the transition line is 3.62 and 6.78,
430 respectively. Because Phimai is a VOC-rich environment, the regime transition occurs at higher R_{FN} values
431 than the conventional definition. This echoes the results of Schroder et al. (2017) on a regionally variable
432 transition region. The definition of $R_{FN} < 1$ as a VOC-limited regime might not be valid in this case.
433 Considering the mean R_{FN} ratio along the transition line (i.e., 6.78), the VOC- and NO_x-limited (and
434 other) regimes are defined as $R_{FN} < 6.78$ and $R_{FN} > 6.78$, respectively. Based on this definition, around 34%
435 (65%) of the ratios are higher (lower) than 6.78, classifying Phimai as a dominant VOC-limited region,



436 which contradicts earlier results (Fig. 5). Biomass burning affects Phimai from January to April and is a
437 significant emission source besides biogenic emission. Thus, high O_3 occurrences likely occur only 30%
438 of the time in a year, and such events are mostly lying above the transition line.

439 In Pantnagar, high O_3 occurrences lie below (42%) and above (57%) the transition line, indicating that O_3
440 production is sensitive to both HCHO and NO_2 , which contradicts Biswas et al. (2019). Pantnagar is a
441 sub-urban site situated beside a busy road. Therefore, the effect of the anthropogenic emissions is
442 expected year-round and the pyrogenic emissions during the spring and post-monsoon period. The O_3
443 sensitivity to both NO_x and VOCs in the northwest IGP region has also been reported in Kumar and Sinha
444 (2021). The mean and minimum R_{FN} value along the transition line is 5.59 and 6.09. The minimum value
445 (i.e., 5.59) is higher than Phimai (3.26), suggesting higher VOC levels in Pantnagar, consistent with the
446 observations.

447 In Chiba, 60% of the R_{FN} values lie below the transition line, suggesting a dominant VOC-limited
448 region, consistent with the results of Irie et al. (2021). The minimum and the mean R_{FN} along the transition
449 line are 0.33 and 0.72, respectively. The transition occurs at a low R_{FN} value due to higher NO_2 levels.
450 Around 40% of the R_{FN} values above the transition region suggest a moderate impact of HCHO on the
451 ozone sensitivity in Chiba.

452 Although the new classification results seem to be reasonable, they should be interpreted with care.
453 Our current understanding of R_{FN} contradicts the classification of rural sites as VOC-limited. Despite the
454 theoretical and observational evidence (i.e., Souri et al., 2020), the classification of regimes based on a
455 single transition line is not yet well-established. Schroder et al. (2017) used regionally varying transition
456 ranges. Moreover, (a) the number of observations and (b) the systematic and retrieval errors can impact
457 the estimations and classifications. These findings are expected to contribute to the ongoing discussion
458 on the effectiveness of R_{FN} . However, the results clarify the idea of a regionally varying transition range.

459
460
461

462 3.1.3.2 R_{FN} profiles



463 Figure 7 shows the seasonal mean R_{FN} profiles at the three sites. The profiles during the high O_3
464 concentrations at the sites (i.e., March in Phimai, May in Pantnagar, and February in Chiba) are only
465 shown. The R_{FN} values will likely increase with height, because of the lower vertical gradient of NO_2 ,
466 compared to HCHO (Fig. 4). Interestingly, the R_{FN} values are almost similar in the 1-2 km height under
467 biomass burning conditions, suggesting a small variability in the HCHO loss rate in the particular layer.
468 At both sites, the HCHO concentration at 1.5 km is about 3 ppbv. In Chiba, a significant amount of NO_2
469 in the higher layers increases the ratio up to 2 km. Above 2 km, the ratio variability at all sites is opposite
470 the surface. The gradient issue of R_{FN} has been discussed explicitly by Jin et al. (2017). They proposed a
471 conversion factor to account for gradient differences in the surface and column-derived R_{FN} values,
472 estimating the conversion factor from the model simulated surface and column abundances of NO_2 and
473 HCHO. We adopt the method reported by Jin et al. (2017) for this study using the CHASER simulated
474 NO_2 and HCHO concentrations and vertical columns.

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

477

$$478 \quad E_{NO_2} = \frac{NO_2 \text{ total column}}{NO_2 \text{ near-surface number density}} \quad (7)$$

479

$$480 \quad E_{HCHO} = \frac{HCHO \text{ total column}}{HCHO \text{ near-surface number density}} \quad (8)$$

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

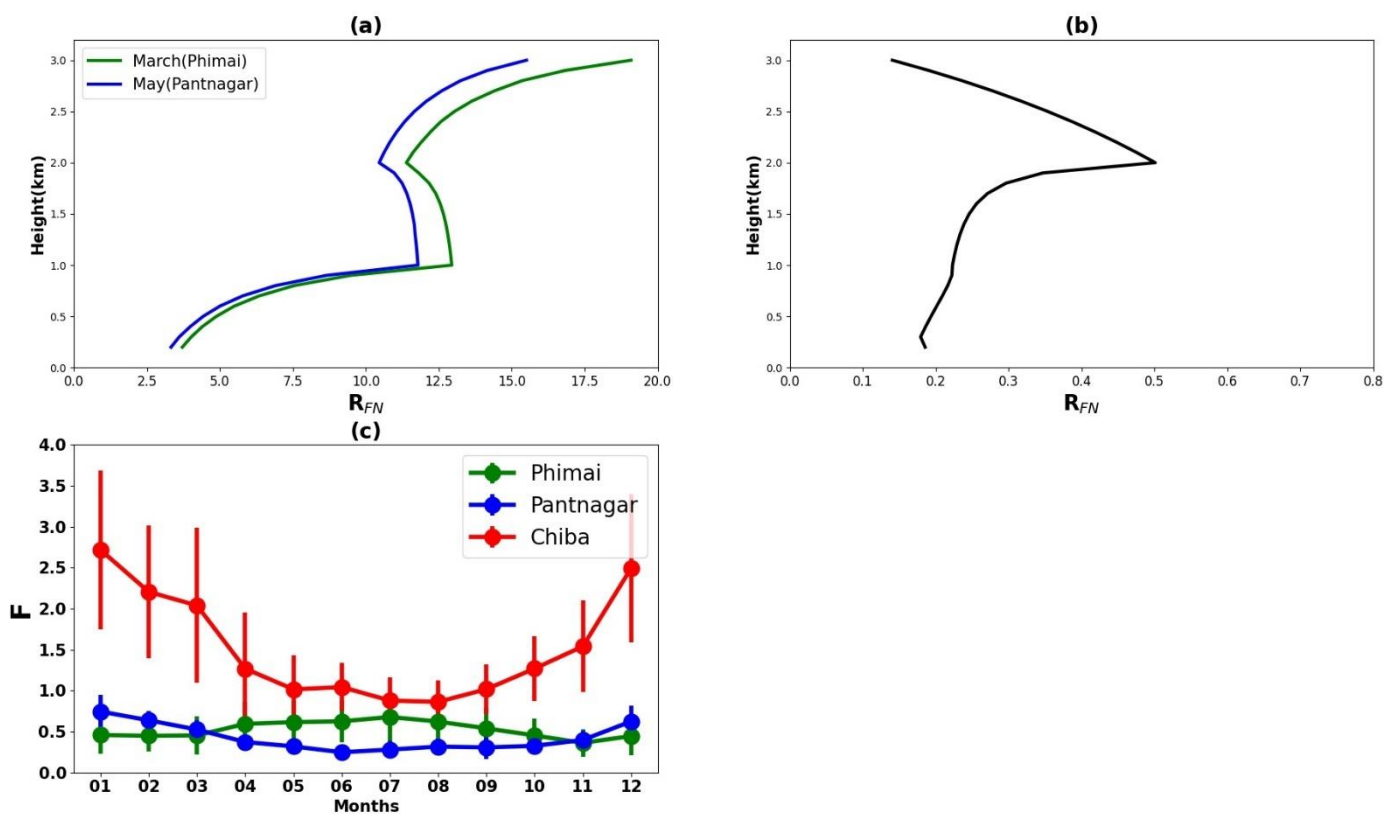
482 In the second step, the column to surface conversion factor (F) was calculated according to the following
483 equation:

$$484 \quad F = \frac{E_{HCHO}}{E_{NO_2}} \quad (9)$$

485 The seasonal variation of F for the three A-SKY sites and the associated 1σ standard deviation of the
486 mean values are depicted in Fig. 7(c). The F values over East Asia reported by Jin et al. (2017) were ~ 2 ,
487 with no marked seasonal variation. Compared to values reported in the literature, the CHASER estimated



488 F values over Chiba range between 1–2.5, which seems reasonable despite the coarse resolution of
489 CHASER. The literature values are reported for polluted regions ($\text{NO}_2 > 2.5$ molecules cm^{-2}) and
490 considered the simulation data for 1–2 PM, whereas the estimates of this study used daytime (07:00 –
491 18:00) simulation.



492

493 **Figure 7:** The seasonal mean R_{FN} profiles during (a) March and May in Phimai and Pantnagar, respectively, and
494 (b) February in Chiba. The months corresponding to the high O_3 concentrations at the respective sites. (c) The
495 seasonal variations in the column to surface conversion factor (F) for the Phimai, Pantnagar, and Chiba sites,
496 estimated from the CHASER simulated HCHO and NO_2 surface concentrations and VCD. The simulated data from
497 07:00 – 18:00 in 2017 were used to estimate the F values. The error bars indicate the one sigma standard deviation
498 of the mean values.

499



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

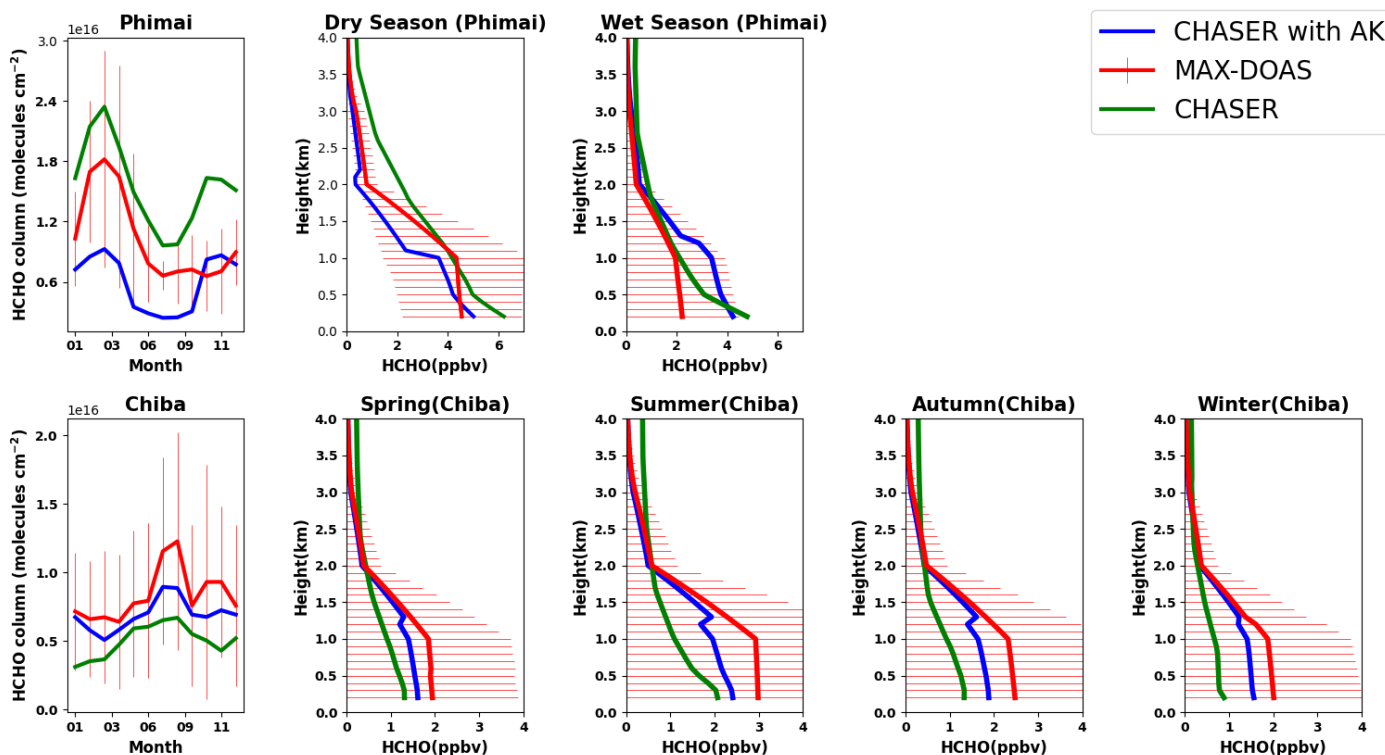
508

509 **3.2 Evaluation of CHASER simulations at the three sites**

510 **3.2.1 Evaluation of CHASER HCHO in Phimai and Chiba**

511 The seasonally averaged observed and modeled HCHO profiles and partial columns in the 0–4 km altitude
512 range in Phimai and Chiba are shown in Fig. 8. The CHASER outputs smoothed with MAX-DOAS
513 averaging kernels (AK) are also depicted. The AK is applied following Franco et al. (2015). First, the
514 CHASER HCHO profiles are interpolated to the MAX-DOAS vertical grids. Next, the MAX-DOAS AK
515 information from individual retrieved profiles is seasonally averaged according to the climate
516 classifications of each site. Finally, the CHASER outputs on the coincident days is selected, and the AK
517 is applied to the daily mean interpolated profile. The coincident days at Phimai and Chiba were 690 and
518 668, respectively. The comparisons at the Pantnagar site are discussed separately.

519 At the Phimai, CHASER predicted the increase in the HCHO partial columns during the dry season
520 and well-reproduced the HCHO seasonality, with an R -value of 0.96. The modeled monthly mean values
521 during the dry season are within the 1σ standard deviation of the observed values, indicating the pyrogenic
522 emissions estimates used for the simulations are reasonable. CHASER predicted a 41% increase in the
523 HCHO column from January to March, consistent with the observations (41%). CHASER overestimates
524 the HCHO columns in both seasons, and the mean bias error (MBE) (CHASER – MAX-DOAS) is lower
525 (3.7×10^{15} molecules cm^{-2}) (Table 5) in the wet season. Although underestimated, the dry season
526 smoothed column values are within the 1σ range.



527 **Figure 8.** The seasonal variations in the HCHO partial columns from 0 to 4 km and vertical profiles during all
528 seasons at Phimai and Chiba, inferred from the MAX-DOAS observations (red) and CHASER simulation (green).
529 The CHASER HCHO partial column and vertical profile smoothed with the MAX-DOAS AK are colored in blue.
530 The AK information of all the screened (as explained in section 2.2) retrievals were averaged based on the seasonal
531 classification of the respective sites. The coincident time and date between the model and observations are selected
532 only. The error bars indicate the one sigma standard deviation of mean values of the MAX-DOAS observations.
533

534 The modeled and observed HCHO concentrations in the 1- 2km layers during the wet season are almost
535 identical, whereas concentrations near the surface (i.e., 0-1 km) differ by 30%. The absolute mean
536 difference in the 0-4 km layer is ~0.45ppbv, with the maximum difference of 2.58 ppbv below 200 m.
537 CHASER has demonstrated good skills in reproducing the HCHO profile in the 0.5 – 4 km layer during
538 the wet season. The significance of AK information is low for the wet season; however, smoothing the
539 model profiles reduces the overall MBE by 43%.

540



541 **Table 5:** Comparison of the seasonal mean HCHO partial columns and profiles (0-4 km) between MAX-DOAS
542 and CHASER at Phimai and Chiba. MBE (CHASER – MAX-DOAS) is the mean bias error. The partial column
543 and profile MBE units are $\times 10^{16}$ molecules cm^{-2} and ppbv, respectively.

Site	Season	Partial column MBE	Smoothed Partial column MBE	Profile MBE	Smoothed Profile MBE
Phimai	Overall	0.28	-0.07	0.35	0.01
Phimai	Dry	0.37	-0.28	0.58	-0.38
Phimai	Wet	0.21	0.07	0.45	0.33
Chiba	Overall	-0.12	-0.05	-0.37	-0.11
Chiba	Spring	-0.07	-0.04	-0.22	-0.12
Chiba	Summer	-0.16	-0.08	-0.45	-0.26
Chiba	Autumn	-0.10	-0.04	-0.40	-0.19
Chiba	Winter	-0.09	-0.01	-0.42	0.11

544

545

546 During the dry season, the absolute mean and maximum difference in the datasets in the 0-1 km layers
547 is ~ 1 ppbv and ~ 2 ppbv, respectively. The observed and simulated seasonal differences in the 0-1 km are
548 50 and 34%, respectively. Simulated dry season profile values above ~ 2 km is out of the 1σ variability
549 range. The two-potential reasons for such differences are lower measurement sensitivity in the free
550 troposphere and overestimated biogenic emissions in the model. Despite the measurement limitations,
551 CHASER and MAX-DOAS wet season profiles up to 3km are consistent. Thus, it is likely that the
552 biogenic emissions in the model are overestimated. However, the dry season HCHO profiles in 0-2 km
553 are well simulated. The smoothing underestimates the dry season profile within the 1σ variability range
554 but improved simulations below 200 m. Above 3 km, the smoothed values mostly imitate the apriori due
555 to reduced measurement sensitivity (i.e., low AK value, indicating limited information could be
556 retrieved).

557

558 A moderate correlation ($R=0.58$) is observed between the modeled and observed HCHO partial
559 columns at Chiba. CHASER could reproduce the peak in the partial columns in August. The model



560 predicts a 41% increase in the HCHO columns from January to August, whereas the observed increase is
561 54%. Although Chiba is an urban site, the HCHO and temperature seasonal variations show a tight
562 correlation ($R \sim 0.70$) (Fig S2), suggesting changes in biogenic emissions modulates HCHO seasonality.
563 Similarly, the modeled seasonality is consistent with temperature variability (Fig. S2). Thus, the simulated
564 HCHO seasonality in Chiba is reasonable, despite underestimated absolute values. Smoothing the
565 simulations improves the correlation, and the MBE is reduced by 54% (Table 4).

566 The CHASER HCHO profiles in the 0- 4 km layers are lower than the observations, with an MBE of
567 0.39 ppbv. The absolute differences in the modeled and retrieved HCHO profiles in the 0-2 km layer
568 during all seasons are higher than in Phimai. Absolute mean differences ~ 1 ppbv and higher mainly are
569 observed from 0 to 2 km. In addition, the vertical gradients of the simulated profiles are low compared to
570 Phimai. The modeled profiles in Chiba resemble the HCHO profiles measured over the ocean during the
571 INTEX-B (Intercontinental Chemical Transport Experiment: Phase B) (Boeke et al., 2011). The Chiba
572 site is near the sea, and coarse CHASER resolution includes the ocean pixels. Moreover, urban surfaces
573 are not homogeneous. Thus, a significant part of the profile discrepancies is likely related to the systematic
574 differences, in addition to emission estimates. However, the model estimates lie within the standard
575 deviation range of the measurements. Due to the low gradients in the simulated profiles, the smoothed
576 profiles mostly imitated the a priori values even below 2 km. The impact of the horizontal resolution on
577 the simulated HCHO levels is discussed in section 3.2.4.

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585 **3.2.2 Evaluation of CHASER NO₂ in Phimai and Chiba**

586 Figure 9 presents the seasonal averages of the MAX-DOAS and CHASER NO₂ profiles and partial
587 columns (0-4 km) in Phimai and Chiba. The AK is applied to the modeled outputs for the Chiba site only.



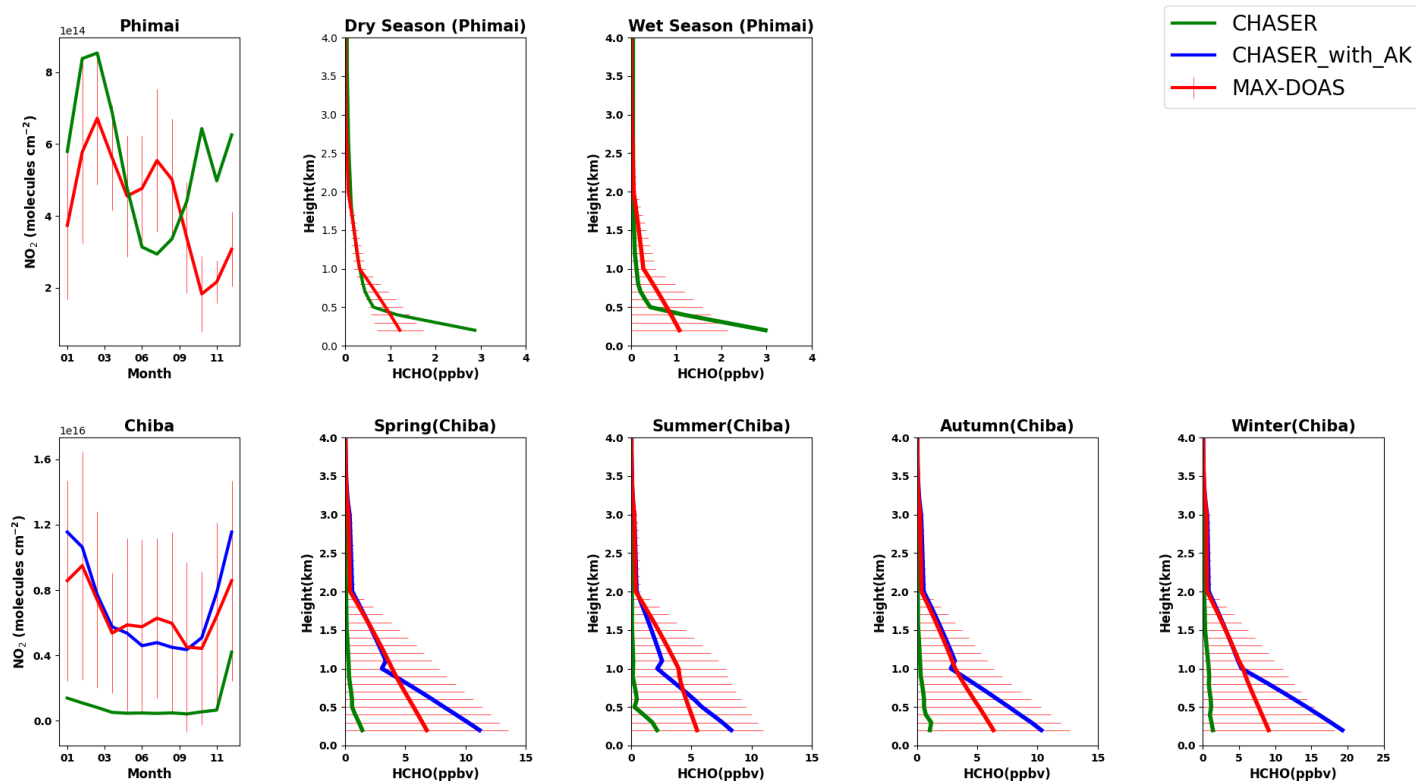
588 Figure S3 in the supplementary information compares the observations, model, and smoothed model
589 profiles averaged within the 0-2 km layer in Phimai. The smoothing with different apriori values is
590 depicted to demonstrate the impact of the apriori values. The smoothed NO₂ concentrations, calculated
591 using the original a priori values, show a seasonal variation shift. The mean smoothed profile resembles
592 the observations when apriori values are reduced by 50%; however, the dry season values are similar in
593 both cases. The apriori values are sensitive to the surface because a priori data are taken from the retrieved
594 VCD and SCD values. Two test cases of smoothing profiles utilizing apriori values above 500 and 800 m
595 shows good agreement with the observations; however, the results are sensitive to the apriori values.
596 Because smoothed profiles are strongly biased to the apriori choice, the smoothing results for the Phimai
597 site are discarded.

598 The modeled NO₂ partial column in Phimai shows good agreement with the observations during the dry
599 season. CHASER well reproduces the enhanced NO₂ columns due to biomass burning within the standard
600 deviation of the observations. The peak in the NO₂ levels during March is consistent in both datasets.
601 Although the seasonality does not agree in other months, the overall MBE is 8×10^{13} molecule cm⁻²
602 (Table 6). Above 500 m, the datasets shows excellent agreement. The absolute mean differences in the 0-
603 1km layer are 0.22 ppbv, and the maximum difference of ~1.9 ppbv is observed near the surface. Amidst
604 the biomass burning influence, the NO₂ concentrations in Phimai are mostly < 1 ppbv. Thus, the
605 comparison results demonstrate CHASER's good skills in a region characterized by low NO₂
606 concentrations. Moreover, when NO₂ concentrations are less than < 1 ppbv, the AK information seems
607 less significant if the model can capture low concentration scenarios.

608



609



610

611 **Figure 9.** The seasonal variations in the NO_2 partial columns from 0 to 4 km and vertical profiles during all seasons
612 at Phimai and Chiba, inferred from the MAX-DOAS observations (red) and CHASER simulation (green). The
613 CHASER NO_2 partial column and vertical profile smoothed with the MAX-DOAS AK are colored in blue. The
614 coincident time and date between the model and observations are selected only. The error bars indicate the one
615 sigma standard deviation of mean values of the MAX-DOAS observations.

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620



621 **Table 6:** Comparison of the seasonal mean NO₂ partial columns and profiles (0-4 km) between MAX-DOAS and
622 CHASER at Phimai and Chiba. MBE (CHASER – MAX-DOAS) is the mean bias error. The partial column and
623 profile MBE units are $\times 10^{15}$ molecules cm⁻² and ppbv, respectively.

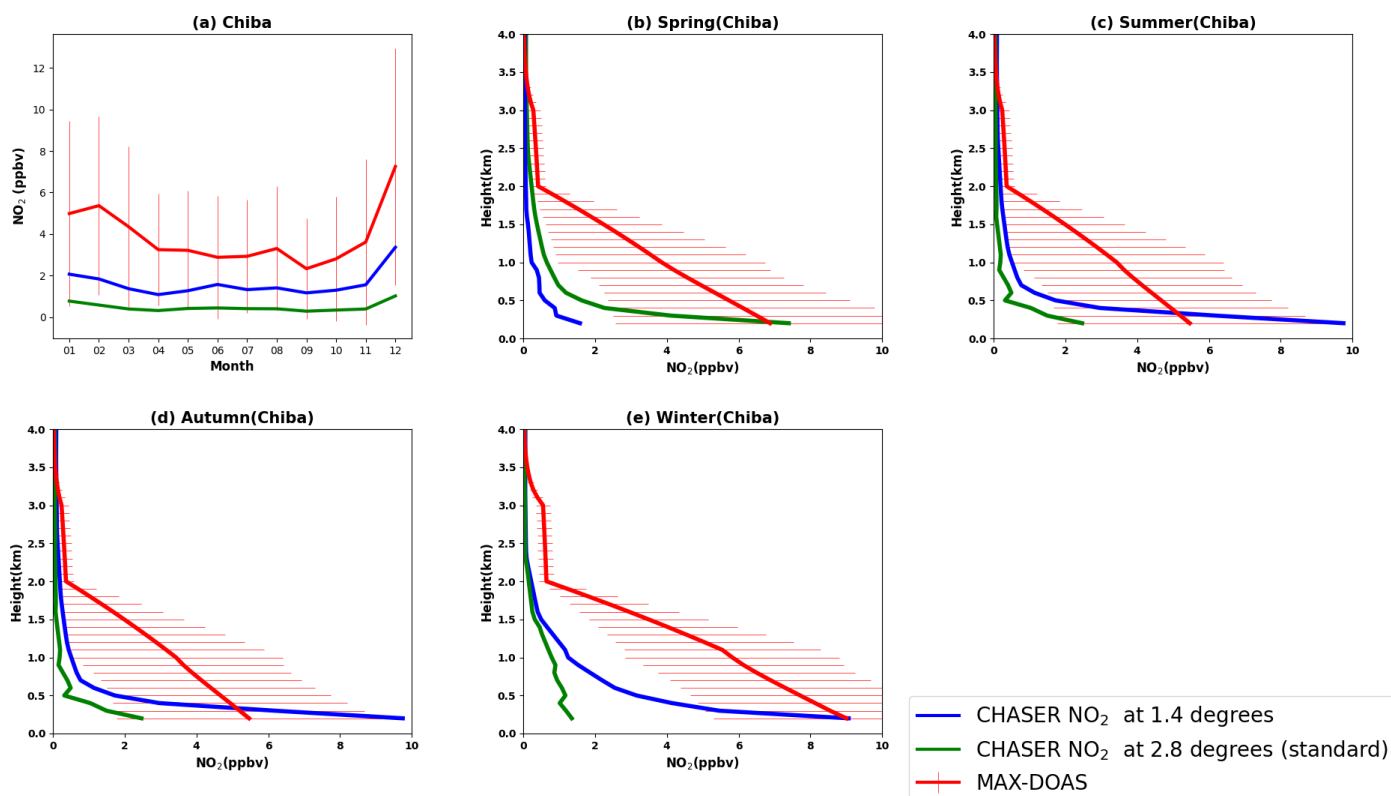
Site	Season	Partial column MBE	Smoothed Partial column MBE	Profile MBE	Smoothed Profile MBE
Phimai	Overall	0.08		0.11	
Phimai	Dry	0.18		0.09	
Phimai	Wet	-0.14		0.02	
Chiba	Overall	-5.58	-1.90	-3.27	-1.66
Chiba	Spring	-5.56	-2.00	-3.19	-1.74
Chiba	Summer	-5.52	-2.87	-2.85	-1.86
Chiba	Autumn	-4.57	-1.24	-2.74	-1.40
Chiba	Winter	-6.64	-1.50	-4.30	-1.63

624
625 Although the datasets are moderately correlated ($R=0.59$) in Chiba, the model largely underestimates the
626 NO₂ partial column with MBE of $\sim 5 \times 10^{15}$ molecules cm⁻². The model predicts almost constant NO₂
627 profiles and columns throughout the year, and thus the respective seasonal biases are almost similar. The
628 vertical gradient of the modeled NO₂ profiles is low, too, similar to the HCHO profiles. The model
629 resolution could be a potential cause of such significant underestimation. The AKs significantly improved
630 the partial column and profiles, reducing the MBE by more than 50%; however, the a priori values strongly
631 affect the smoothed profiles.

632
633 The mean NO₂ concentrations in the 0-2 km layer in 2018, simulated at the spatial resolution of $2.8^\circ \times$
634 2.8° (standard) and $1.4^\circ \times 1.4^\circ$ are compared with observations in Chiba, depicted in Fig.10(a). The error
635 bars are the 1σ standard deviation of the observations. Higher resolution simulations reduced the overall
636 MBE by 35% (Table 7). NO₂ concentrations at 1.4° are now within the variability of the observations.
637 The 1.4° simulation captured the NO₂ seasonal variability better than at 2.8° . Despite an improved
638 resolution, the model values are underestimated, with the highest MBE during the winter. According to
639 Miyazaki et al. (2020), the seasonality in the anthropogenic emissions, primarily wintertime heating, is



640 not included in the emission inventories, which could likely underestimate the wintertime NO_2 levels.
641 The best agreement between the datasets is observed during summer and spring, with an MBE of ~ 1 ppbv
642 on a seasonal scale.
643



644 **Figure 10:** (a) The seasonal variations in the NO_2 concentrations in the 0-2 km layer in Chiba, inferred from the
645 MAX-DOAS observations (red) and two CHASER simulations at 2.8° (green) and 1.4° (blue) resolutions. The
646 simulated NO_2 profiles at 2.8° (green) and 1.4° (blue) resolutions during (b) spring, (c) summer, (d) autumn, and (e)
647 winter are plotted with the observed seasonal profiles in Chiba. Only the data (both observed and simulated) for
648 2018 are plotted. The coincident time and date between the model and observations are selected only. The error
649 bars in (a), (b), (c), and (d) indicate the one sigma standard deviation of mean values of the MAX-DOAS
650 observations.

651

652



653

654 NO₂ profiles at 2.8° and 1.4° resolution are evaluated in Fig. 10(b-e). The significant impact of the
655 increased resolution is observed below 500 m, reducing the negative bias by 70% near the surface. Above
656 500 m, the effect of the higher resolution is limited, with an MBE reduction of 12% in the 0.6 – 2 km.
657 Although the near-surface NO₂ concentrations at 1.4° resolution are overestimated, the values are within
658 the standard deviation of the observations. Around 200m, the wintertime mean NO₂ concentrations at 1.4°
659 resolution are identical to the observations (~9 ppbv), and the summertime mean is overestimated.
660 Moreover, the NO₂ levels above 2km are similar at both resolutions. The resolution effect on NO₂ profiles
661 varies with location and season (Williams et al., 2017). For example, CHASER NO₂ at 1.1° resolution
662 significantly improved the agreement with aircraft observations below 650 hPa over Denver metropolitan
663 area (Sekiya et al. 2018), whereas, in Chiba, the 1.4° resolution improved the surface estimates. Thus, the
664 horizontal resolution is not the only reason for the model underestimation. Other factors such as the
665 vertical resolution, uncertainties in emission inventories, chemical kinetics, etc., can also affect the
666 simulated NO₂ estimates. The impact of the emission inventory is discussed in section 3.2.4.

667

668 **Table 7:** Comparison of the seasonal mean NO₂ profiles (0-2 km) among MAX-DOAS and CHASER simulations
669 at 2.8° and 1.4° resolutions at Chiba. MBE at (CHASER – MAX-DOAS) 1.4° and 2.8° are the mean bias error at
670 the respective resolutions. The MBE unit is ppbv.

Season	MBE at 1.4°	MBE at 2.8°
Overall	-2.24	-3.37
Spring	-2.26	-3.23
Summer	-1.50	-2.47
Autumn	-1.57	-2.57
Winter	-3.44	-5.07

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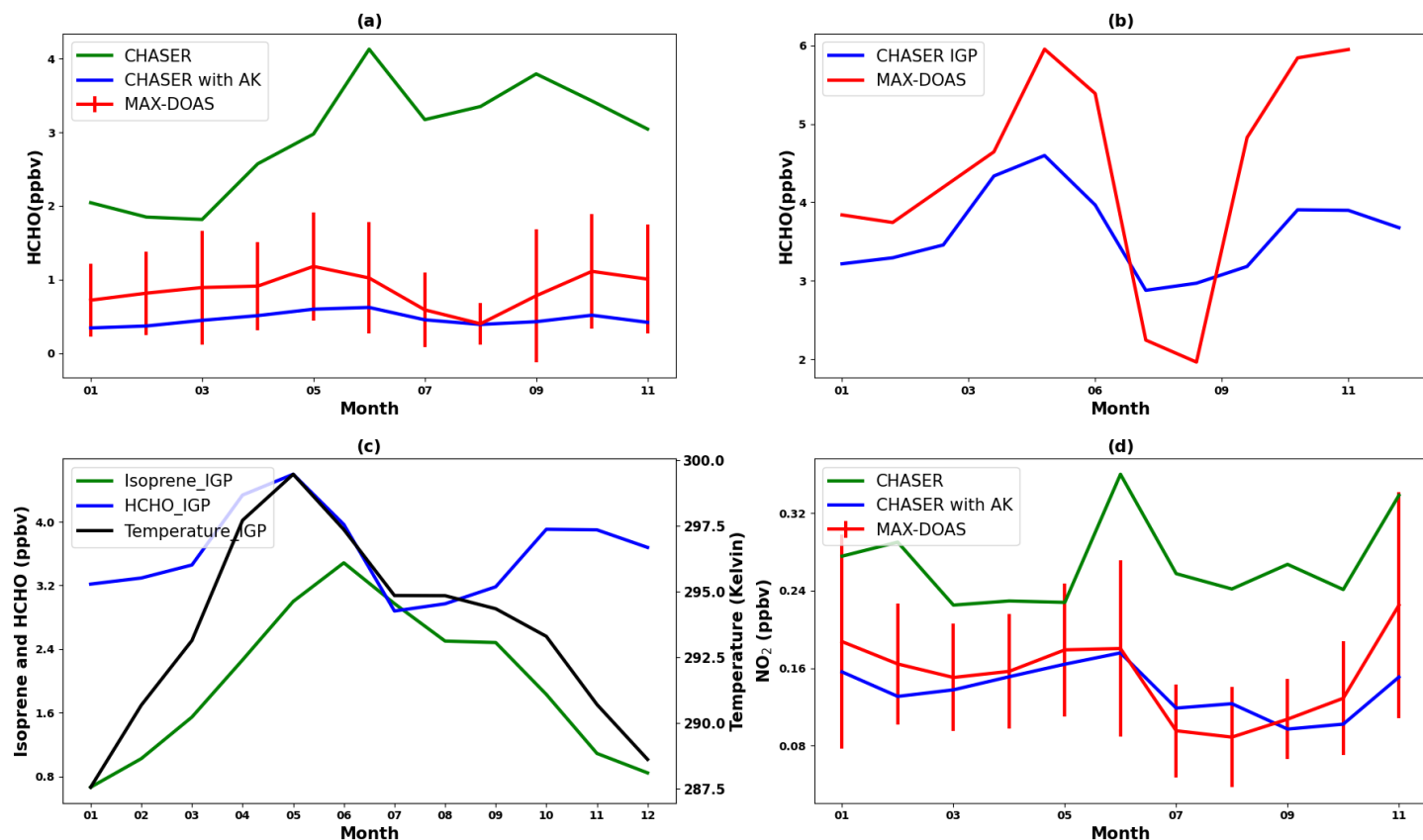
673



674 **3.2.3 Evaluation of CHASER HCHO and NO₂ in Pantnagar**

675 Mountainous terrain is located a few kilometers away from the Pantnagar site. The treatment of such
676 complex landscapes is different in the datasets. The simulated surface height at Pantnagar is 1.7 km,
677 whereas the radiative transfer calculations for MAX-DOAS retrievals consider a flat terrain. Such an
678 assumption is expected to have a non-significant effect on the retrieval because the horizontal resolution
679 of the MAX-DOAS system is ~10 km. Moreover, the MAX-DOAS HCHO products in Pantnagar showed
680 good agreement with satellite observations (Hoque et al., 2018b). Because the MAX-DOAS observations
681 are most sensitive between 0-2 km, the observations above 1.8 km are compared with the simulations.
682 The observed, modeled, and smoothed mean HCHO levels in the 1.8 – 3 km layer are shown in Fig. 11
683 (a). The datasets differ significantly with an MBE of ~2.5ppbv, and simulated values are out of the 1σ
684 variability range. Smoothing the model output reduced the MBE to 0.65 ppbv, which correlates well with
685 the observation with an *R*-value of 0.66. Besides the systematic differences in the datasets, two potential
686 factors can contribute to the significant discrepancies in the datasets- (1) similar to Phimai, the simulated
687 HCHO levels in Pantnagar are higher than observations, indicating overestimated biogenic emissions. (2)
688 Restricting the observation ELs to less than 10° reduces the sensitivity to the higher layers, affecting the
689 concentration estimates in the upper layers.

690 Figure 11(b) compares the modeled HCHO simulations over the whole IGP region with the observations
691 in Pantnagar. The observed HCHO seasonality in Pantnagar is consistent with Mahajan et al. (2015),
692 reported for the entire IGP region. Thus, the comparison can assess the model capability in the IGP region.
693 The comparison for 2017 is shown only because the observations were continuous and covered almost
694 all the seasons.



695

696 **Figure 11.** (top panel) (a) The seasonal variations in the HCHO concentrations in the 1.8–3 km layer in Pantnagar,
697 inferred from the MAX-DOAS observations (red) and CHASER simulation (green). The simulated concentrations
698 smoothed with the MAX-DOAS AK are shown in blue. (b) The seasonal variations in the MAX-DOAS (red) and
699 CHASER (blue) HCHO concentrations in Pantnagar and the IGP region, respectively, in 2017. The coincident
700 dates between the observations and model are plotted only. (bottom panel) (c) The CHASER simulated seasonal
701 variations in HCHO (blue), isoprene (green), and temperature (black) in the IGP region in 2017. Only the daytime
702 simulated values were considered for the plot. (d) The observed (red) and modelled (green) seasonal variations in
703 the NO₂ concentrations in the 1.8–2 km layer in Pantnagar. The simulated concentrations smoothed with the MAX-
704 DOAS AK are shown in blue. The error bars in (a) and (d) are the one sigma standard deviation of the observations.

705

706



707 The modeled HCHO seasonal variations in the IGP region correlate well with $R \sim 0.80$. The
708 enhancement in the HCHO concentrations during the spring and post-monsoon season is well reproduced
709 by CHASER. This indicates that CHASER can capture HCHO variability in complex terrain regions like
710 IGP. Figure 11(c) depicts the isoprene concentrations and temperature in the IGP region, in addition to
711 the HCHO concentrations. Oxidization of precursor hydrocarbon and photochemical reactions are the
712 most dominant sources of HCHO. Isoprene is the most abundant hydrocarbon in the atmosphere. The
713 average ambient isoprene concentrations during July, August, and September in the IGP region are
714 1.4 ± 0.3 ppbv (Mishra et al., 2020). Therefore, the CHASER isoprene concentration range of 1.5–2 ppbv
715 during the monsoon season seems reasonable. The HCHO concentrations in the IGP region reach a peak
716 during the spring and post-monsoon seasons. A strong correlation between HCHO, isoprene, and
717 temperature variation ($R \sim 0.90$) in the first half of the year indicates that the change in biogenic emissions
718 strongly drives the HCHO seasonal modulation. The observed enhancement in the HCHO levels during
719 spring in Pantnagar is related to biomass burning. The biomass burning events are primarily concentrated
720 in the northwest IGP region (Kumar and Sinha, 2021), where the site is located. On a regional scale, the
721 biomass burning impact is expected to smear. Thus, the strong effect of the biogenic emission on the
722 regional HCHO modulation is reasonable. HCHO modulation differs from isoprene and temperature
723 during the post-monsoon period, suggesting a more substantial role of biomass burning and anthropogenic
724 emissions. Thus, the physical processes driving the HCHO seasonality in the IGP region are well reflected
725 in the CHASER simulations.

726 The observed, modeled, and smoothed mean NO_2 levels in the 1.8 – 2 km layer are shown in Fig. 11 (d).
727 The layers 1.8–2 km are considered instead of 1.8 – 3 km because of the lower gradient of NO_2 than
728 HCHO (Fig. 4). Despite overestimated modeled values, the datasets are moderately correlated with R
729 ~ 0.52 . Smoothing the model output increases the correlation to 0.74. The model estimated seasonal
730 modulation is 40%, whereas the observed modulation is 60%, with the maximum absolute difference of
731 ~ 0.18 ppbv. CHASER could capture the NO_2 variability (< 1 ppbv) in Pantnagar near the free troposphere.
732 According to the datasets, the mean NO_2 concentrations range from 0.2 to 0.35 ppbv at 2 km in Pantnagar
733 during biomass burning events. The simulated seasonal NO_2 estimates near the free troposphere (1.8 – 2
734 km) are consistent with the observations at all three sites.

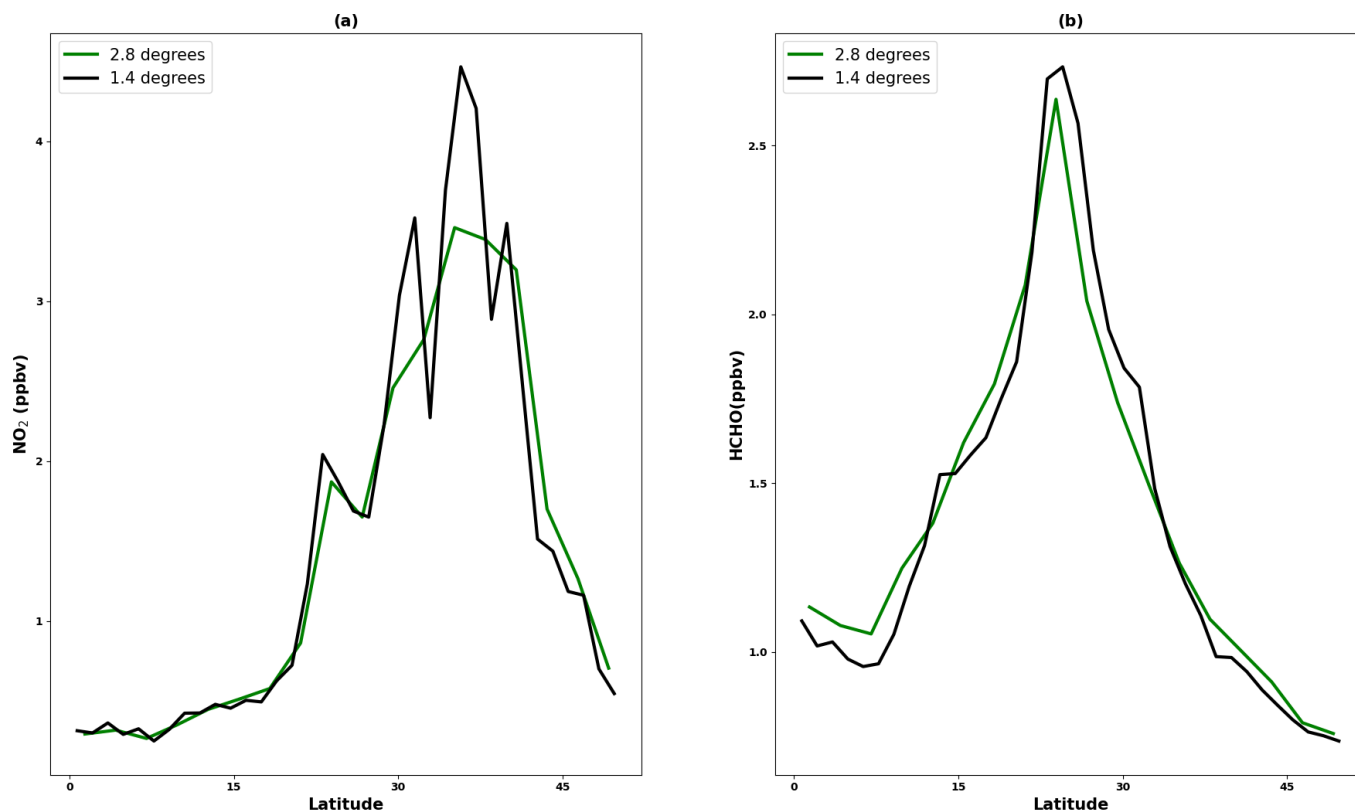


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736 **3.2.4 Impact of the model resolution and emission inventories on the results**

737 The impact of the spatial resolution on the evaluation results is assessed by comparing the CHASER
738 simulations at 2.8° and 1.4° resolutions, shown in Fig. 12. The latitude and longitude range of the averaged
739 values is 0°-50° N and 60° – 150° E, respectively, covering the location of all three sites. Simulations for
740 2018 are plotted and analysed only. The two NO₂ simulations differ on average by 5%. The differences
741 in the latitude range of 28° – 40° N and 10° – 20°N is around 10 and 3 %, respectively. This suggests that
742 the NO₂ comparison results in South and East Asia will improve by at least 10% at a higher resolution.
743 The discussion in section 3.2.2 showed that the resolution-based improvement in the overall profile was
744 limited and varied regionally. At a spatial resolution of 1.1°, CHASER NO₂ columns can capture the
745 regional NO₂ climatology over polluted and biomass burning region better than at 2.8° (Sekiya et al.
746 2018). The biomass burning enhanced NO₂ levels were well reproduced in Phimai at 2.8°; however, the
747 concentrations were less than 1 ppbv. NO₂ simulations at 2.8° were reasonable in low NO₂ regions. The
748 HCHO levels in Chiba and the rest of the region improved by around 8 and 2 %, respectively, indicating
749 low sensitivity to spatial resolution. Therefore, HCHO estimates at 1.4° resolution would yet be
750 underestimated in Chiba. Overall, the impact of the model resolution on the results is estimated to be
751 around 20%.

752 Although the NO_x estimates for the low NO₂ regions seem reasonable, global NO_x emissions have
753 changed since 2008(i.e., EDGAR-HTAP (2008) emissions used in this study). A recent study by Miyazaki
754 et al. (2020) reported changes in global NO_x emissions from 2005 to 2018. They found a continuous 30%
755 increase in NO_x emissions in India since 2005 and a substantial increase in Southeast and Southern Asia.
756 Thus, the current simulation settings will likely underestimate the NO₂ concentrations in these regions.
757 Figure S4 depicts the simulated and measured NO₂ concentrations in two polluted cities (Delhi and
758 Kolkata) in India. Updated NO_x estimates would yield higher NO₂ concentrations and improve the
759 comparison, yet likely to be underestimated.



760

761

762 **Figure 12.** The zonal mean (a) NO₂, and (b) HCHO concentrations were estimated from CHASER simulations at
763 the spatial resolution of 2.8° × 2.8° (green) and 1.4° × 1.4° (black). The latitude and longitude bound of the mean
764 values are 0° - 50°N and 60° - 150°E, respectively. The simulations for 2018 and daytime values from 09:00-
765 15:00 LT are plotted.

766

767 Chiba is not regarded as a major polluted region. Irie et al. (2021) reported a declining trend in NO₂ levels
768 in Chiba since 2012, echoing Miyazaki et al. (2020)'s results over Japan. The current model setting with
769 updated NO_x estimates would simulate a lower NO₂ level in Chiba than the observations. Therefore, the
770 impact of an updated emission inventories will also vary regionally. Data assimilation techniques can be
771 adopted to improve the model-observation comparison, which will be discussed in our future studies.

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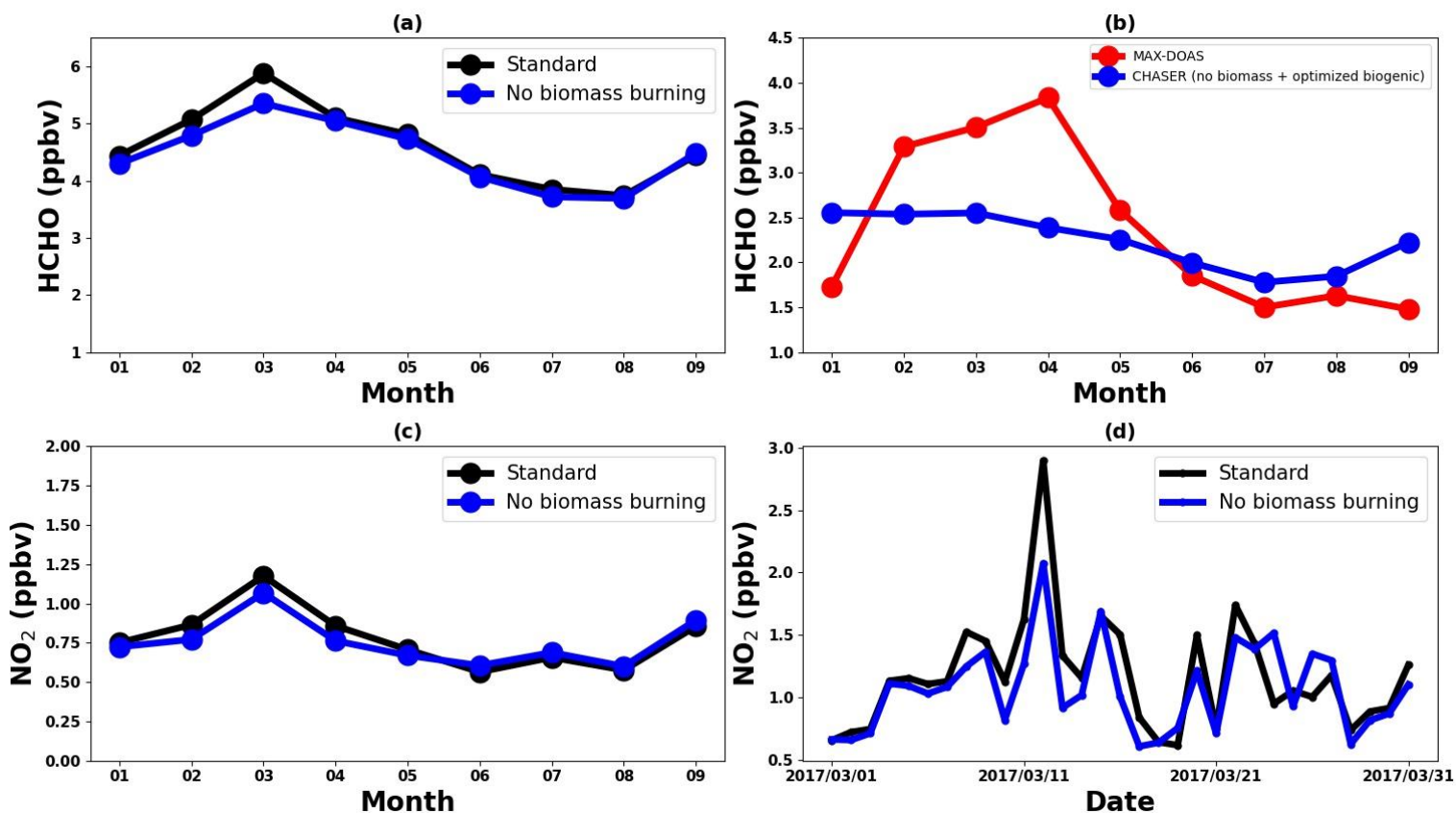
773 3.3 Contribution estimates

774 3.3.1 Contribution from biomass burning to the HCHO and NO₂ abundances in Phimai

775 The good agreements between the datasets in the 0-1 km layer in Phimai can quantify biomass burning
776 contributions to the HCHO and NO₂ concentrations. Figure 13 presents the results of simulation
777 L1_HCHO, L1_opt, and L1_NO₂. The simulation settings are shown in Table 2. For better readability,
778 the switched-off emissions criterion is described in the legends of Fig. 13. The plots present the mean
779 concentrations in the 0–1 km layer. Biomass burning contributes ~10% to the HCHO concentrations in
780 Phimai during the dry season. However, based on the observations, a more substantial effect of biomass
781 burning is expected. During the wet season, the MAX-DOAS and CHASER HCHO concentrations are,
782 respectively, ~2 and ~4 ppbv, indicating overestimation of the biogenic emissions in CHASER. Figure
783 13(b) shows the HCHO concentration obtained from simulation L1_opt and MAX-DOAS observations
784 in 2017. In the L1_opt simulation setting, the biomass burning emissions are switched off; the biogenic
785 emissions are optimized to reproduce results analogous to those obtained from observations during the
786 wet season. In the absence of biomass burning, the surface HCHO concentrations in Phimai would be ~2
787 ppbv, resulting in a biomass burning contribution of ~20–50% during the dry season. The observed
788 interseason difference in the HCHO concentration in Phimai is ~60%. Consequently, the revised biomass
789 burning contribution estimate is more reasonable. Pyrogenic emissions contributions to the NO₂
790 concentrations in Phimai are ~10% during the dry season (Fig. 13(c)). Because the NO₂ concentrations
791 are low in Phimai, the simulation during March, when the influence of biomass burning is highest, is used
792 to derive a better contribution estimate. In the absence of biomass burning, the NO₂ concentration during
793 March would be ~0.84 ppbv (Fig.13(d)), indicating a contribution up to 35% to the NO₂ concentrations
794 in Phimai.



795



796

797 **Figure 13.** (top panel) The seasonal variations in the HCHO concentrations in the 0-1 km layer in Phimai, obtained
798 from the standard and L1_HCHO simulations. The pyrogenic emissions of VOCs are switched off in L1_HCHO.
799 (b) The HCHO seasonal variability in Phimai in 2017, obtained from the MAX-DOAS observations (red) and
800 L1_opt simulations. The pyrogenic VOC emissions were switched off, and the biogenic emissions were reduced
801 by 50% in L1_opt. The coincident dates between the observation and the simulations are plotted only. (bottom
802 panel) (c) The seasonal variations in the NO₂ surface concentrations in Phimai in 2017, obtained from the standard
803 and L1_NO₂ simulations. (d) Standard and L1_NO₂ simulation outputs of the daily mean NO₂ surface
804 concentrations during March 2017. The pyrogenic NO₂ emissions were switched off in the L1_NO₂ simulation.
805 Only the daytime values from 09:00 – 15:00 LT are used to calculate the seasonal mean.

806

807



808 **3.3.2 Contribution of soil NO_x emissions in Phimai**

809 Because soil NO_x emissions are included in CHASER simulations, the NO₂ contributions from soil
810 emissions are quantified. Figure 14 presents the monthly mean surface NO₂ concentrations in Phimai in
811 2017, simulated including (standard) and switching off (L1_NO₂) the soil NO_x emissions. The NO₂
812 concentrations between 09 and 12 hr. were used to calculate the monthly mean concentrations. Soil
813 emissions contribute ~20% of the overall NO₂ concentrations in Phimai, with higher contributions during
814 the wet season. The highest soil contribution of about 25% occurs in July.

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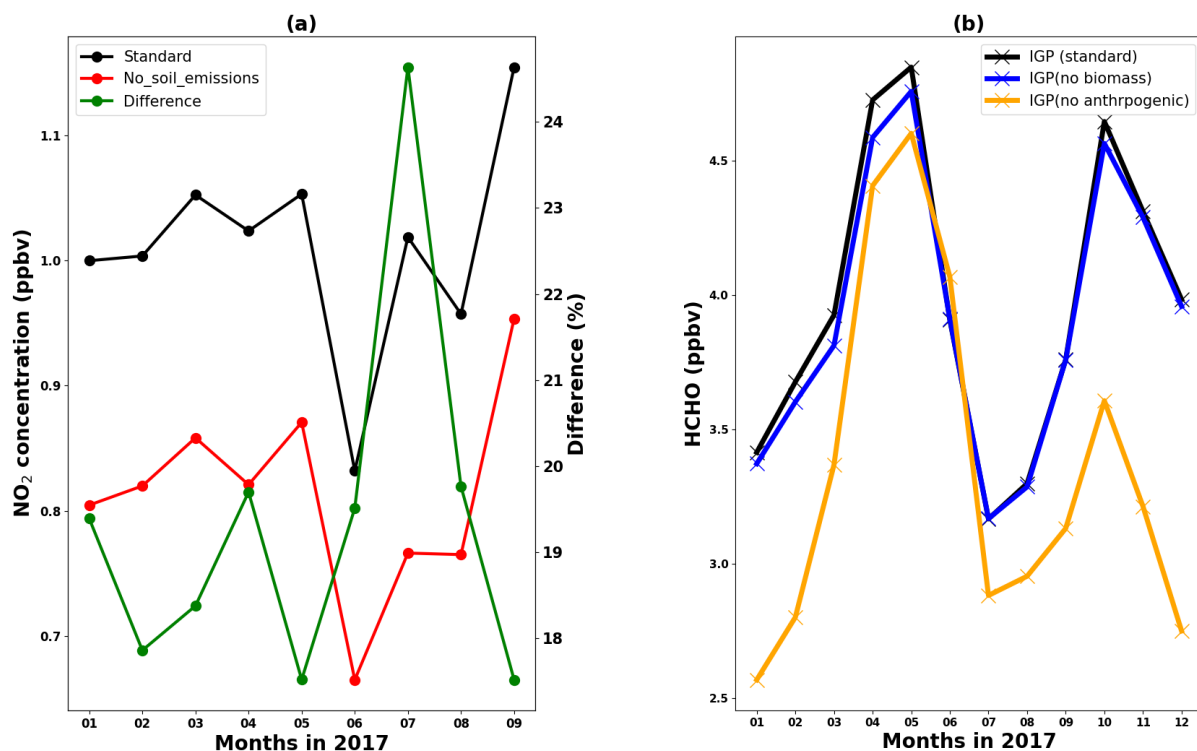
816 **3.3.3 Contribution from pyrogenic and anthropogenic emissions to the HCHO abundances in the** 817 **IGP region**

818 Figure 14(b) presents the standard, L1_HCHO (pyrogenic VOC emissions switched off), and L2
819 (anthropogenic VOC emissions switched off) HCHO simulations in the IGP region. According to
820 L1_HCHO simulation results, the effect of biomass burning emissions on the regional HCHO modulation
821 is small (~12%). The HCHO concentrations in India have biogenic, anthropogenic, and pyrogenic VOC
822 sources. However, Biogenic VOCs are the primary driver of the over HCHO variation (Surl et al., 2018).
823 Consequently, two potential reasons might be responsible for the small effects of pyrogenic emissions on
824 HCHO concentrations: (1) Overestimation of the biogenic emission or underestimation of pyrogenic
825 emissions in the model. (2) Stronger effects of anthropogenic VOC emissions than pyrogenic VOCs. The
826 L2 simulations show that anthropogenic emissions contribute up to 30% of the HCHO concentration in
827 the IGP region, with a maximum contributed during the post-monsoon season, which coincides with the
828 lower isoprene concentration (i.e., biogenic emissions) and temperature (Fig. 12(c)). Moreover, Kumar
829 and Sinha (2021) reported high acetaldehyde concentrations from anthropogenic emissions in the IGP
830 region throughout the year. Thus, anthropogenic emissions are likely to be a significant driver of HCHO
831 concentrations in the IGP region after biogenic emissions.

832



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834

835 **Figure 14.** (a) The monthly mean NO₂ concentrations in Phimai were estimated from the standard (black) and
836 L1_NO₂ (red) simulations. The soil NO_x emissions are switched off in the L1_NO₂ simulation. The green line
837 indicates the percentage difference between the two simulations. (b) The seasonal variations in the HCHO
838 concentrations in the IGP region, obtained from the standard, L1_HCHO (pyrogenic VOC emission switched off),
839 and L2 simulations (anthropogenic VOC emissions switched off). The simulations for 2017 are plotted and
840 analyzed. Daytime values from 09:00-12:00 and 09:00 – 15:00 LT were selected for Phimai and IGP, respectively.

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846 **4 Conclusions**

847 NO₂ and HCHO concentrations and profiles are retrieved from MAX-DOAS observations at three A-
848 SKY sites from January 2017 through December 2018, using the JM2 algorithm. The retrieved products
849 were used to evaluate the global chemistry transport model CHASER simulations at the three sites. At all
850 three locations, the seasonal variation of both trace gases was consistent throughout the investigated
851 period. At Phimai and Pantnagar, biomass burning led to enhanced HCHO and NO₂ concentrations,
852 respectively, during the dry season and spring and post-monsoon season. At Chiba, the HCHO variability
853 was consistent with the temperature-led seasonal changes in the biogenic emissions. The changes in the
854 dry season HCHO and NO₂ levels in Phimai from 2015 to 2018 were consistent with the number of fire
855 events.

856 The R_{FN} values were biased towards a particular regime when the standard transition range $1 < R_{FN} < 2$
857 was used. The parameterization of Souri et al. (2020) provides a better estimate of the transition region.
858 The classification results of the revised transition region in Phimai and Pantnagar contradicted the results
859 based on the standard transition range; however, they were more reasonable. Such a method is based on
860 observations, thus influenced by measurement constraints. More observational evidences are required to
861 standardize this method. Overall, the results clarified that the standard transition region is not valid
862 globally.

863 CHASER showed good skills in Phimai, characterized as a VOC-rich and low NO₂ (<1 ppbv) region. In
864 both seasons, the observed and modeled profiles (HCHO and NO₂) agreed within the one sigma standard
865 deviation of the measurements, despite a general overestimation of the model. Furthermore, both the wet
866 season HCHO profiles were almost identical in the 0.5 – 4 km layer in both datasets.

867 At Pantnagar, the data within the 1.8 – 2 km layer were evaluated due to the mountainous terrain of the
868 site. The simulated NO₂ concentrations around 2 km were less than 1 ppbv, moderately correlated with
869 the observations with a maximum absolute difference ~0.2 ppbv. The model overestimated the HCHO
870 level at the same height.

871 CHASER demonstrated limited performances in Chiba. NO₂ at higher resolution (i.e., 1.4°) mainly
872 improved the surface estimates, reducing the overall MBE in the 0-2 km layer by 35%. The finer



873 resolution would improve the HCHO estimates in Chiba by 10%; however, it is yet to be underestimated.
874 Sensitivity studies for the Phimai site estimated biomass burning contributions to the HCHO and NO₂
875 concentrations up to ~50 and ~ 35%, respectively. On average, 20% of the NO₂ level originates from soil
876 NO_x emissions, which increased to 25% in July. Anthropogenic emissions (contribution up to 30%) have
877 a more substantial impact on the VOC variability in the IGP region than biomass burning, consistent with
878 reports in the literature.

879

880

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

885

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

889

890 *Author contributions:* HMSH conceptualized the study, conducted the model simulations, analyzed the
891 data, and drafted the manuscript. AMF helped with the data processing. HI developed the JM2 code and
892 maintained the A-SKY network. KS developed the CHASER model and supervised the study. MN is the
893 PI of the Pantnagar site. AD and MN shared their experience to explain the results. HI, KS, AD, MN, and
894 AMF commented and provided feedback on the final results and manuscript.

895

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

897

898 *Acknowledgments:* This research is supported by the Global Environmental Research fund (S-12 and S-
899 20) of the Ministry of the Environment (MOE), Japan, and JSPS KAKENHI Grants: JP20H04320,
900 JP19H05669, and JP19H04235. The CHASER model simulations are partly performed with the



901 supercomputer (NEC SX-Aurora TSUBASA) at the National Institute for environmental studies (NIES),
902 Tsukuba, Japan.

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