



1 Glyoxal tropospheric column retrievals from TROPOMI, 2 multi-satellite intercomparison and ground-based validation

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24 **Abstract.** We present the first global glyoxal (CHOCHO) tropospheric column product derived from the
25 TROPOspheric Monitoring Instrument (TROPOMI) on board of the Sentinel-5 Precursor satellite. Atmospheric
26 glyoxal results from the oxidation of other non-methane volatile organic compounds (NMVOCs) and from direct
27 emissions caused by combustion processes. Therefore, this product is a useful indicator of VOC emissions. It is
28 generated with an improved version of the BIRA-IASB scientific retrieval algorithm relying on the Differential
29 Optical Absorption Spectroscopy (DOAS) approach. Among the algorithmic updates, the DOAS fit now includes
30 corrections to mitigate the impact of spectral misfits caused by scene brightness inhomogeneity and strong NO₂
31 absorption. The product comes along with a full error characterization, which allows providing random and
32 systematic error estimates for every observation. Systematic errors are typically in the range of 1-3x10¹⁴
33 molec/cm² (~30-70% in emission regimes). Random errors are larger (>6x10¹⁴ molec/cm²) but can be reduced by
34 averaging observations in space and/or time. Benefiting from a high signal-to-noise ratio and a large number of
35 small-size observations, TROPOMI provides glyoxal tropospheric column fields with an unprecedented level of
36 details.

37 Using the same retrieval algorithmic baseline, glyoxal column data sets are also generated from the Ozone
38 Monitoring Instrument (OMI) on Aura and from the Global Ozone Monitoring Experiment-2 (GOME-2) on board
39 of Metop-A and Metop-B. Those four data sets are intercompared over large-scale regions worldwide and show
40 a high level of consistency. The satellite glyoxal columns are also compared to glyoxal columns retrieved from
41 ground-based Multi-Axis (MAX-) DOAS instruments at nine stations in Asia and Europe. In general, the satellite
42 and MAX-DOAS instruments provide consistent glyoxal columns both in terms of absolute values and variability.
43 Correlation coefficients between TROPOMI and MAX-DOAS glyoxal columns range between 0.61 and 0.87.



44 The correlation is only poorer at one mid-latitude station, where satellite data appears low biased during
45 wintertime. The mean absolute glyoxal columns from satellite and MAX-DOAS generally agree well for
46 low/moderate columns with differences less than 1×10^{14} molec/cm². A larger bias is identified at two sites where
47 the MAX-DOAS columns are very large. Despite this systematic bias, the consistency of the satellite and MAX-
48 DOAS glyoxal seasonal variability is excellent.

49 1. Introduction

50 Exposure to poor air quality kills millions of people annually (e.g. Vohra et al., 2021; World Health Organization,
51 2016) due to natural and human emissions of a large range of particulate matters and gases, including among
52 others nitrous oxides (NO_x), sulphur dioxide, carbon monoxide, methane and volatile organic compounds
53 (VOCs). The latter, in combination with NO_x, play a significant role in the secondary production of tropospheric
54 ozone (Jacob, 2000), which is highly toxic for the respiratory system and also contributes to global warming
55 because of its absorption in the thermal infrared. Global measurements of atmospheric concentrations of the ozone
56 precursors is therefore crucial. The number of VOCs that can be found in the atmosphere is manifold, but only a
57 few of them can be probed using remote sensing techniques. For example, formaldehyde (HCHO) measurements
58 have been used in many studies as a proxy for probing emissions of non-methane VOCs of biogenic, pyrogenic
59 and anthropogenic origin (e.g. Abbot et al., 2003; Barkley et al., 2013; Bauwens et al., 2016; Beekmann and
60 Vautard, 2010; Curci et al., 2010; Jin et al., 2020; Marais et al., 2012; Palmer et al., 2006; Stavrakou et al., 2016;
61 Wells et al., 2020).

62 With a lifetime of a few hours, glyoxal (CHOCHO) is another short-lived VOC that can be detected remotely,
63 offering the potential to provide information on Non-Methane VOC (NMVOC) emissions. Over the past few
64 years, an increasing number of studies (e.g. Cao et al., 2018; Chan Miller et al., 2017; Fu et al., 2008; Li et al.,
65 2016; Liu et al., 2012; Stavrakou et al., 2009, 2016; Wittrock et al., 2006) have exploited glyoxal measurements
66 from space, often in combination with formaldehyde. Being produced from similar sources, those two species are
67 complementary as they have different production yields. For example, the oxidation of aromatics produces glyoxal
68 with a much higher yield than formaldehyde (Cao et al., 2018). Although being both mostly produced via the
69 oxidation of other VOCs, direct emissions from anthropogenic and fire activities also occur, and contribute more
70 to the glyoxal global budget than to the formaldehyde one (Stavrakou et al., 2009b, 2009a). This motivated many
71 studies to investigate the ratio of glyoxal to formaldehyde concentrations or columns as a possible metric to
72 discriminate between different types of VOC emissions (e.g. Chan Miller et al., 2014; DiGangi et al., 2012; Hoque
73 et al., 2018; Kaiser et al., 2015; Vrekoussis et al., 2010). Glyoxal measurements are also essential for establishing
74 the global budget of secondary organic aerosols (SOAs). Indeed, with a high solubility in water, glyoxal undergoes
75 heterogeneous uptake on aerosols and cloud droplets where the subsequent aqueous-phase chemistry forms SOA
76 (Chan et al., 2010; Fu et al., 2008; Hallquist et al., 2009; Knote et al., 2014; Li et al., 2016; Volkamer et al., 2007).

77 Glyoxal has three absorption bands in the visible spectral range that have been exploited to remotely retrieve
78 information on its atmospheric abundance using the Differential Optical Absorption Spectroscopy method
79 (DOAS, Platt and Stutz, 2008) applied to ground-based (e.g. Benavent et al., 2019; Hoque et al., 2018; Javed et
80 al., 2019; Schreier et al., 2020), air-borne (e.g. Kluge et al., 2020; Volkamer et al., 2015), ship-borne (e.g. Behrens
81 et al., 2019; Sinreich et al., 2010) and space-based instruments. The first global glyoxal tropospheric column



82 observations from space have been realized by Wittrock et al. (2006) using nadir measurements from the
83 SCIAMACHY (SCanning Imaging Absorption spectroMeter for Atmospheric CartographY) instrument. Based
84 on this pioneering work, different glyoxal data products were derived from the Global Ozone Monitoring
85 Experiment-2 (GOME-2) (Lerot et al., 2010; Vrekoussis et al., 2009) and from the Ozone Monitoring Instrument
86 (OMI) (Alvarado et al., 2014; Chan Miller et al., 2014). All those different products rely on a similar DOAS
87 approach, but generally differ from each other by the choice of the fit settings and of the auxiliary input data.

88 In general, the glyoxal optical depth is very low ($< 5e-4$), typically one order of magnitude smaller than the NO_2
89 optical depth in the same spectral range. This results in retrievals prone-to-noise, requiring to average many of
90 them to extract meaningful glyoxal signals. With an enhanced spatial resolution resulting in a number of
91 observations more than ten times larger than provided by its predecessor OMI, the TROPospheric Monitoring
92 Instrument (TROPOMI), operating since 2017, allows observing weak atmospheric absorbers with an
93 unprecedented level of spatio-temporal details. This has been illustrated by Alvarado et al. (2020a) who
94 investigated the large amounts of formaldehyde and glyoxal emitted by the intense North-American wildfires in
95 August 2018 as observed by TROPOMI for several days and over long distances. Theys et al. (2020) have
96 evaluated the respective contributions to the hydroxyl radical production in fresh fire plumes from nitrous acid,
97 VOCs and other sources with the support of different TROPOMI data sets, including the glyoxal data product
98 described here.

99 This work presents the latest version of the BIRA-IASB scientific glyoxal tropospheric column retrieval algorithm
100 that has been applied to three years of TROPOMI measurements, and also to data from the predecessor nadir
101 instruments OMI and GOME-2A/B. The quality of the TROPOMI glyoxal retrievals is investigated with (1) a
102 global intercomparison of the satellite glyoxal data products generated with a common algorithm and (2)
103 comparisons with independent glyoxal measurements from a series of Multi-AXis DOAS (MAX-DOAS)
104 instruments located at nine stations in Asia and Europe.

105 After a brief introduction of the satellite instruments used in this study in Section 2, the retrieval algorithm and its
106 different steps are described in Section 3, with emphasis on the updated and innovative aspects compared to
107 heritage studies. This section also presents the typical random and systematic errors associated to the retrievals
108 and how they are estimated for each individual measurement. Section 4 presents the evaluation of the inter-satellite
109 consistency by comparing both seasonal global spatial patterns as seen from different instruments as well as
110 monthly mean time series and seasonal cycles in a series of selected large-scale regions. Finally, Section 5 presents
111 validation results based on MAX-DOAS data.

112 2. TROPOMI and other nadir-viewing satellite sensors

113 TROPOMI was launched on 13 October 2017 on board of the Sentinel-5 precursor platform. It flies on a sun-
114 synchronous Low Earth Orbit (LEO) with an ascending node crossing the equator at the local time of 13:30. In
115 the series of Sentinel missions from the European Union Copernicus programme, it is the first one dedicated to
116 atmospheric composition. The instrument operates in a nadir viewing mode and measures Earthshine radiances
117 and solar irradiances in the ultraviolet (UV), visible, near infrared and short infrared spectral bands. It aims at
118 providing column amounts of a number of key pollutants, such as ozone (O_3), NO_2 , SO_2 , HCHO, CO, CH_4 as well
119 as cloud and aerosol parameters. TROPOMI offers a quasi-daily global coverage at the unprecedented spatial



120 resolution of $3.5 \times 5.5 \text{ km}^2$ ($3.5 \times 7 \text{ km}^2$ before August 2019) in the UV-visible spectral range. It is an imager-type
121 instrument using a two-dimensional Charge Coupled Device (CCD) for the light measurements, the detector
122 columns being used for the spectral resolution while the rows are binned to resolve spatially the 2600 km across-
123 track swath into 450 individual ground pixels. The spectral resolution of the instrument is about 0.5 nm and offers
124 a remarkably high signal-to-noise ratio of about 1500 in band 4 (405-500 nm) used in this study. More details on
125 the instrument and its performance can be found in (Kleipool et al., 2018; Ludewig et al., 2020; Schenkeveld et
126 al., 2017; Veeffkind et al., 2012). The TROPOMI measurements allow to derive the vertical columns of multiple
127 species, some of them not included among the operational products listed above. Glyoxal is one of them and the
128 details on how its column quantities are retrieved will be described in the next section.

129 The TROPOMI design strongly inherits from past nadir-viewing sensors, and in particular from the Ozone
130 Monitoring Instrument (OMI) that we use to evaluate the TROPOMI glyoxal product presented in this work. OMI
131 (Levelt et al., 2006) is also an imager instrument and flies on an early afternoon orbit since October 2004. The
132 OMI swath, divided into 60 across-track pixels with a size varying from $13 \times 24 \text{ km}^2$ (at nadir) to $13 \times 150 \text{ km}^2$ (at
133 the edges), allowed a daily global coverage before being limited in 2008 by the so-called row anomaly. The latter
134 consists in a modification of the signal recorded by OMI at specific rows, due to a mechanical obstruction of the
135 field of view, and leads to lower quality spectral measurements (Torres et al., 2018). We also exploit spectral
136 measurements from the Global Ozone Monitoring Experiment-2 (GOME-2) instruments aboard the Metop-A and
137 Metop-B platforms. In contrast to OMI and TROPOMI, the GOME-2 instruments (Munro et al., 2016) fly on
138 early morning LEOs with local equator crossing times around 09:30 and are scanning spectrometers, meaning that
139 across-track pixels are successively sounded. The scan is divided into 24 pixels for a total swath of 1920 km,
140 providing global coverage in 1.5 day. Each pixel has a size of $80 \times 40 \text{ km}^2$. After the launch of Metop-B, the GOME-
141 2A swath was reduced to 960 km in July 2013, leading to ground pixel two times smaller.

142 3. Description of the Algorithm

143 The algorithm for retrieving tropospheric vertical columns of glyoxal relies on a classical DOAS approach (Platt
144 and Stutz, 2008). This approach consists first in fitting measured optical depths in an optimized spectral window
145 to derive the so-called slant column densities *SCDs* (atmospheric concentration integrated along the effective light
146 path) of the absorbers. The latter are thereafter converted into vertical column densities *VCDs* (concentration
147 vertically integrated from the satellite ground pixel up to the top of the atmosphere) with air mass factors obtained
148 by modelling the radiative transfer through the atmosphere. An additional background correction procedure is
149 often applied for weak absorbers such as glyoxal in order to reduce as much as possible the presence of systematic
150 biases caused by spectral interferences.

151 The glyoxal algorithm presented here largely inherits from past developments for predecessor nadir-viewing
152 satellite sensors (Alvarado et al., 2014, 2020; Chan Miller et al., 2014; Lerot et al., 2010; Vrekoussis et al., 2009;
153 Wittrock et al., 2006). In the following subsections, we further describe each algorithmic component, with
154 emphasis on its specificities. The retrievals are provided with estimates for the random and systematic errors,
155 which are discussed in subsection 3.4.



156 **3.1. DOAS fit**

157 To exploit the glyoxal absorption bands, we use a fitting window from 435 to 460 nm encompassing the two most
158 intense bands, which has shown in the past to provide reliable results (Barkley et al., 2017; Lerot et al., 2010).
159 This has been confirmed by sensitivity tests carried out by Alvarado et al. (2014) and Chan Miller et al. (2014).
160 Owing to its low optical depth ($<5 \times 10^{-4}$), any poorly fitted feature in the radiance measurements may affect the
161 retrieved glyoxal SCD. It is therefore crucial to account for any physical or instrumental effect in order to optimise
162 the fit quality as much as possible. Different aspects of the algorithm contribute to achieve this.

163 The wavelength grids of the measured spectra are recalibrated before the actual DOAS fits with a cross-correlation
164 procedure (Danckaert et al., 2017; De Smedt et al., 2018) during which the position of the lines in the measured
165 irradiance spectrum is fitted to an external solar atlas (Chance and Kurucz, 2010), convolved to the satellite
166 spectral resolution. This recalibration procedure is done once per orbit and separately for every detector row of
167 the instrument.

168 Although the DOAS fit generally uses an irradiance as the reference spectrum, it is common practice, in the case
169 of weak tropospheric absorbers, to replace it by a mean radiance spectrum recorded in a remote region where the
170 concentration of the gas of interest is low (e.g. De Smedt et al., 2018). This allows reducing the presence of
171 systematic biases caused by spectral interferences and/or instrumental limitations. In particular, the use of one
172 separate mean radiance spectrum per detector row minimizes the presence of so-called stripes in the product
173 typical of imager-type instruments such as OMI or TROPOMI. Here we compute those mean radiance spectra on
174 a daily basis by averaging for each row all spectra located within the equatorial Pacific Ocean (15°S-15°N;
175 120°W-180°W).

176 The selected settings for the DOAS fits rely on the aforementioned past studies and are summarized in Table 1.
177 The latest available cross-sections for species absorbing in the selected fitting window are included in the fit, i.e.
178 O₃, NO₂, O₂-O₂, water vapour and liquid water in addition to glyoxal. Note that the water vapour cross-section is
179 based on the HITRAN2012 database (Rothman et al., 2013) as we found that the latest HITRAN2016 version
180 (Gordon et al., 2017) led to poorer fit quality. The temperature dependence of the NO₂ absorption is taken into
181 account by including a second cross-section, taken as the difference between NO₂ cross-sections reported at 2
182 temperatures (220 and 294K) as proposed by Alvarado et al. (2014) and Chan Miller et al. (2014) for their
183 respective OMI glyoxal products. Consistently with Alvarado et al. (2014), we found that fitting the liquid water
184 optical depth in the glyoxal fitting window performs as well as fixing it to a value previously determined in a
185 larger spectral interval as proposed in the past (Lerot et al., 2010). A number of additional cross-sections are
186 included in the fit to consider (1) Inelastic scattering (Ring effect) introduces high-frequency structures that are
187 treated as a pseudo-absorber (Chance and Spurr, 1997); (2) Intensity offsets in the spectra, caused for example by
188 residual straylight, are corrected for by fitting the inverse of the reference spectrum (Danckaert et al., 2017); (3)
189 heterogeneity of the scene brightness may also introduce high frequency structures, which are considered with
190 pseudo-cross-sections (more details hereafter). All those cross-sections are generated at the instrumental spectral
191 resolution by using the key data Instrumental Spectral Response Functions provided for all individual detector
192 rows. During the DOAS procedure, the earthshine radiance spectrum is further aligned with the reference, by
193 allowing it to be shifted and stretched in wavelength. In addition, the DOAS fit procedure includes a spike removal



194 scheme as described in Richter et al. (2011) enabling to filter out from the fit individual corrupted radiance
 195 measurements, and hence to reduce the noise in the product.

196 **Table 1 : Absorption cross-sections and settings used for the retrieval of glyoxal slant columns**

Fitting interval	435-460 nm
Absorption cross-sections	
Glyoxal	Volkamer et al. (2005)
Ozone	Serdyuchenko et al. (2014), 223K
NO ₂	Vandaele et al. (1998), 220K and 294K, I ₀ effect-corrected (Aliwell et al., 2002)
O ₄ (O ₂ -O ₂)	Thalman and Volkamer (2013), 293K
H ₂ O (vapour)	Rothman et al. (2013), 293K
H ₂ O (liquid)	Mason et al. (2016)
Scene Heterogeneity	2 pseudo-absorbers (Richter, 2018) – Internally generated
Ring effect	Pseudo-absorber (Chance and Spurr, 1997; Wagner et al., 2009)
Other parameters	
Polynomial	3rd order
Intensity offset correction	1 st -order offset (additional cross-section taken as the inverse of the reference spectrum)
Earthshine wavelength shift	1 st -order shift
Reference spectrum (E₀)	Daily average of radiances, per detector row, selected in equatorial Pacific (Lat: [-15° 15°], Long: [180°-240°])

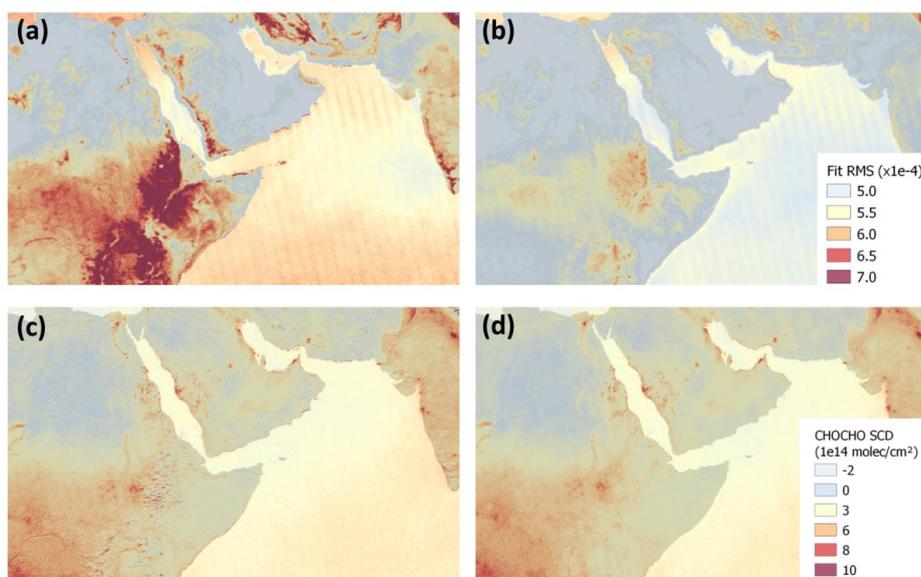
197

198 **3.1.1. Scene heterogeneity**

199 Any intensity variation within the probed scene taking place perpendicularly to the instrumental slit (i.e. along
 200 track) leads to perturbations of the instrumental spectral slit function (ISRF) (Noël et al., 2012; Voors et al., 2006).
 201 Richter et al. (2018) have shown that those perturbations lead to a degradation of the NO₂ DOAS spectral fit quality
 202 and to systematic biases on the retrieved slant columns. Such abrupt intensity changes occur for example along
 203 the coasts, mountains or cloud edges. Glyoxal retrievals are also affected by such scene heterogeneity as illustrated
 204 in Figure 1 over the Horn of Africa and Middle East. This figure shows in the panel (a) that the root mean square
 205 (RMS) of the DOAS fit residuals is systematically higher along the coasts but also over land where contamination
 206 by broken clouds or abrupt elevation changes cause discontinuities in brightness fields. The stripes visible in this
 207 figure are due to the smaller pixel size (and hence lower signal-to-noise ratio) on the edges of the across-track
 208 field of view. The panel (c) shows that there are some collocated artificial patterns (positive/negative biases) in
 209 the mean retrieved glyoxal slant column field. The latter result from spectral interferences with the signature
 210 introduced by the ISRF distortion. Richter et al. (2018) showed that those spectral interferences can be
 211 significantly reduced with additional cross-sections in the DOAS fit scaling the possible scene heterogeneity
 212 signature. Those cross-sections are generated with a statistical analysis of the fit residuals for many observations



213 in a remote region as a function of the level of scene heterogeneity. The latter can be computed using radiance
214 measurements at higher spatial resolution available in the TROPOMI level-1 data at a limited number of
215 wavelengths. Following this approach, two additional cross-sections have been added to the DOAS baseline and
216 both the fit residuals and the identified glyoxal biases have been reduced as illustrated in the right panels (b) and
217 (d) of Figure 1. This effect is particularly visible along coasts and mountains but also over lands where some
218 pseudo-noise caused by persistent broken clouds is also largely reduced. Note that a third cross-section derived
219 from the mean residuals of homogeneous scenes is also added, which explains why the fit RMS are also reduced
220 (but less drastically) in homogeneous scenes. This cross-section has no impact on the retrieved glyoxal SCDs and
221 allowed mostly isolating systematic residuals due to scene heterogeneity only for the pseudo cross-sections
222 creation.



223

224 **Figure 1 : Impact of scene brightness heterogeneity on glyoxal retrievals in the fitting window 435-460 nm over the**
225 **Horn of Africa and Middle-East. The panels (a) and (b) show mean fit residuals RMS for the year 2019 without and**
226 **with (left and right) pseudo-cross sections to correct for spectral signatures introduced by scene heterogeneity. The**
227 **panels (c) and (d) show the corresponding mean glyoxal slant column densities. Only observations with cloud**
228 **fractions less than 20% are considered.**

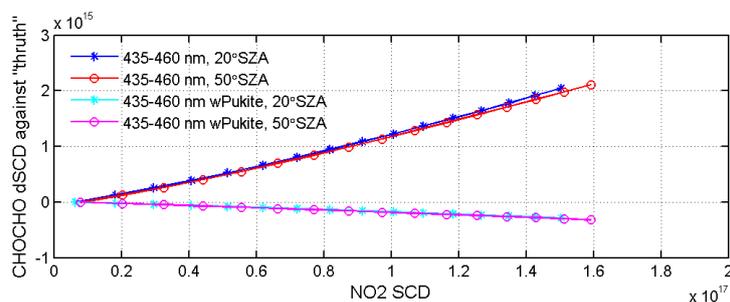
229 3.1.2. Empirical correction for strong NO₂ absorption

230 The DOAS approach assumes that the wavelength dependence of the effective light path within the fit interval
231 can be neglected. Although this assumption is generally reasonable, it may fail in case of strong absorption by one
232 (or more) species, of which the slant column density becomes dependent on the wavelength (Puķīte et al., 2010).
233 In that case, fitting the optical depth of that species by a simple scaling of its cross-section is inaccurate and the
234 fit quality is degraded. Puķīte et al. (2010) have shown that fitting additional cross-sections resulting from a Taylor



235 expansion of the wavelength-dependent slant column corrects for its variability within the fit window. As
236 mentioned before, the high sensitivity of glyoxal retrievals to potential sources of misfit was a motivation to
237 further investigate its sensitivity to extreme NO_2 concentration levels.

238 For this purpose, synthetic spectra were generated at a spectral resolution of 0.5 nm with the radiative transfer
239 model SCIATRAN (Rožanov et al., 2005) for a satellite nadir-viewing geometry and two different solar zenith
240 angles. In those simulations, inelastic scattering was neglected and a large range of tropospheric NO_2 columns
241 was covered by scaling the NO_2 a priori profile. The TROPOMI DOAS baseline described above was then applied
242 to those simulated spectra in order to retrieve CHOCHO SCDs and evaluate the error as a function of the NO_2
243 SCD as illustrated in Figure 2. Results clearly point to a CHOCHO SCD error increasing with the NO_2 SCD. Note
244 that the exact error magnitude may change slightly depending on the NO_2 vertical distribution and on the actual
245 atmospheric content. On the other hand, adding the so-called Pukite cross-sections (Puķīte et al., 2010) to account
246 for the wavelength-dependence of the NO_2 SCD significantly reduces the errors.



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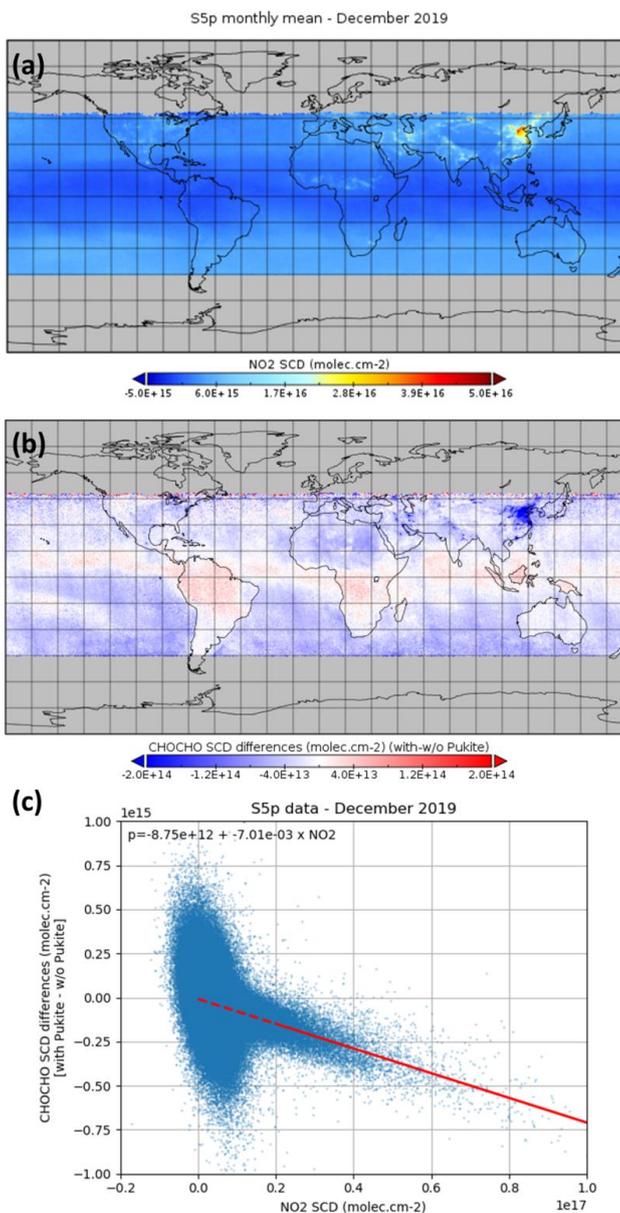
248 **Figure 2 : Absolute error (in molec/cm²) on the retrieved CHOCHO SCD as a function of the NO_2 SCD for simulated**
249 **spectra in a nadir-viewing satellite geometry and for two solar zenith angles. The reference “true” CHOCHO SCD is**
250 **taken as the value retrieved for the lowest NO_2 SCD scenario. The error increases with the NO_2 SCD when Pukite**
251 **cross-sections are not included in the fit, but remains small otherwise.**

252 On this basis, the impact of adding the Pukite cross-sections to the DOAS baseline has been investigated using
253 one month of TROPOMI data. A wintertime period was chosen (December 2019) to favour the number of
254 observations with large NO_2 concentrations, in particular in China but also in other megacities in the Northern
255 Hemisphere. Figure 3 (upper panel) displays the monthly mean NO_2 SCDs in December 2019, and (middle panel)
256 the mean impact on the retrieved CHOCHO SCDs of introducing the Pukite terms in the DOAS spectral fit
257 baseline. The CHOCHO SCD differences caused by the Pukite terms are also plotted as a function of the NO_2
258 SCDs to better visualize the correlation (lower panel). For regions with enhanced NO_2 concentrations ($>2 \times 10^{16}$
259 molec/cm²) (e.g. China, India, Teheran), the Pukite cross-sections lead to a systematic reduction of the CHOCHO
260 SCDs, consistent with the closed-loop tests described above. A small improvement of the fit quality is found (not
261 shown). Unexpectedly, the impact of those additional cross-sections on the CHOCHO SCDs can also be non-
262 negligible in regions with low NO_2 columns: positive differences are for example observed over equatorial oceans,
263 but also over South America and Africa. The correlation plot of Figure 3 clearly shows these two regimes. While
264 the impact of the Pukite cross-sections on the glyoxal retrievals is understood and reliable for large NO_2 SCDs,



265 their influence at low NO₂ SCD is more questionable and likely results from spectral interferences occurring
266 between the different fitted spectra (e.g. with the Ring signature), which introduces additional noise in the product.

267 To avoid this, rather than fitting additional cross-sections, we introduce an empirical correction applied to the
268 glyoxal SCDs. This correction consists in subtracting from the glyoxal SCD an NO₂-SCD dependent value,
269 directly prescribed from the linear regression fit through the sensitivity test results for all observations worldwide
270 from December 2019, with NO₂ SCDs larger than 2×10^{16} molec/cm² as illustrated in Figure 3 (c). It is worth
271 noting that the regression fit results agree well with the glyoxal SCD errors estimated from the simulations
272 presented above (Figure 2). For extreme pollution conditions such as what can be found in China during
273 Wintertime, this correction may lead to glyoxal column reduction up to 30%.



274

275 **Figure 3 : (a) Monthly mean NO₂ SCDs retrieved from TROPOMI data in December 2019. Panel (b) illustrates the**
276 **CHOCHO SCD absolute differences (molec/cm²) due to the incorporation of the Pukite et al. (2010) cross-sections in**
277 **the DOAS spectral fit and panel (c) shows the correlation between those differences and the NO₂ SCDs. The red line**
278 **corresponds to a linear regression fit through all points with NO₂ SCD larger than 2x10¹⁶ molec/cm².**



279 **3.2. Air Mass Factor computation**

280 The computation of the air mass factor (AMF) used to convert the retrieved glyoxal slant columns (SCD) to
281 vertical columns (VCD) relies on the formulation of Palmer et al. (2001), which decouples the radiative transfer
282 through the atmosphere from the vertical distribution of the gas of interest. Radiative transfer simulations are
283 performed with the vector model VLIDORT at the middle of the fitting window (448 nm) to compute so-called
284 altitude-dependent air mass factors or box-AMFs representing the sensitivity of the slant column to a small
285 concentration change at any altitude. The AMF is obtained as the weighted mean of those box-AMFs using as
286 weights the vertical distribution of the glyoxal concentration.

287 Typically, the sensitivity of nadir-viewing UV-Visible instruments is reduced in the lowermost atmospheric layers
288 because of Rayleigh scattering. However, this sensitivity depends strongly on the observation geometry, on the
289 surface reflectivity and altitude and on the presence of clouds. For example, the sensitivity is generally further
290 reduced for low sun elevation. For this reason, retrievals with solar zenith angles larger than 70° are filtered out.
291 We use a pre-computed five-dimensional look-up table of Box-AMFs spanning all observation conditions (see
292 Table 2) and from which appropriate values are linearly interpolated for every TROPOMI observation. This
293 interpolation uses as input the observation angles provided in the level-1 data, surface elevation taken from the
294 GMTED2010 topography (Danielson and Gesch, 2011) and surface albedo extracted from the OMI minimum
295 Lambertian Equivalent Reflectivity climatology (Kleipool et al., 2008). The spatial resolution of the latter
296 database (0.5°x0.5°) is coarse compared to the TROPOMI footprint and neglects anisotropy, which may introduce
297 significant errors (Lorente et al., 2018). However, at the time of writing, it is the only database available at the
298 S5p overpass time although new Lambertian Equivalent Reflectivity climatologies relying on past works (e.g.
299 Loyola et al., 2020; Tilstra et al., 2021, 2017) are currently being prepared. On the other hand, the level of noise
300 in glyoxal retrievals generally requires averaging in space and/or time which in turn will reduce part of those error
301 sources. We also neglect the impact of clouds and aerosols on the radiative transfer. Instead we apply a stringent
302 cloud filtering approach: only observations with an effective cloud fraction (as retrieved in the same spectral range
303 and provided in the TROPOMI operational NO₂ product (van Geffen et al., 2019) lower than 20% are conserved.
304 This approach is motivated by the fact that glyoxal slant columns tend to be biased high over bright scenes because
305 of poorly understood residual spectral interferences (e.g. with the Ring signature). Similarly, scenes covered by
306 snow and ice are also discarded.

307

308

Table 2 : Granularity of the Box-AMF look-up table

Parameter name	Grid of values
Solar zenith angle [deg]	0, 10, 20, 30, 40, 45, 50, 55, 60, 65, 70, 72, 74, 76, 78, 80, 85
Line of sight zenith angle [deg]	0, 10, 20, 30, 40, 50, 60, 65, 70, 75



Relative azimuth angle [deg]	0, 45, 90, 135, 180
Surface albedo	0, 0.01, 0.025, 0.05, 0.075, 0.1, 0.15, 0.2, 0.25, 0.3 0.4, 0.6, 0.8, 1.0
Surface pressure [hPa]	1063.10, 1037.90, 1013.30, 989.28, 965.83, 920.58, 876.98, 834.99, 795.01, 701.21, 616.60, 540.48, 411.05, 308.00, 226.99, 165.79, 121.11

309

310 The a priori glyoxal vertical profile shapes necessary to perform the AMF computations are provided by the global
311 Chemistry Transport Model MAGRITTE developed at BIRA-IASB, which inherits from the IMAGES model
312 (Bauwens et al., 2016; Müller and Brasseur, 1995; Stavrakou et al., 2009b, 2013). This model runs at $1^\circ \times 1^\circ$
313 resolution and calculates the distribution of 182 chemical compounds, of which 141 species undergo transport.
314 The modelled troposphere is vertically divided in 40 levels between the surface and the lower stratosphere and
315 meteorological fields are provided by the ECMWF ERA-5 analyses. The chemical mechanism and deposition
316 scheme have been recently updated (Müller et al., 2018, 2019). Anthropogenic NMVOC emissions of are
317 provided by the EDGAR 4.3.2 inventory (Huang et al., 2017) for the year 2012. Biomass burning emissions are
318 obtained from the Global Fire Emission Database version 4 (GFED4s) (Van Der Werf et al., 2017). The emissions
319 of isoprene and monoterpenes are calculated using the MEGAN-MOHYCAN model (Guenther et al., 2012;
320 Müller et al., 2008). The model also incorporates biogenic emissions of methanol, methyl-butanol, ethylene,
321 ethanol, acetaldehyde, formaldehyde and acetone, as well as oceanic emissions of methanol, acetone, acetaldehyde
322 and alkyl nitrates (Müller et al., 2019). The global source of glyoxal in the model amounts to 47 Tg/yr (in 2013),
323 of which about 4 Tg/yr are due to direct biomass burning emissions, and 18, 6, 9 and 9 Tg/yr are due to the
324 atmospheric degradation of isoprene, acetylene, aromatics and monoterpenes, respectively (Müller et al., 2019).

325 To account for the difference in spatial resolution between the model and the observations, a priori profiles are
326 rescaled to the effective satellite pixel surface elevation using the formulation proposed by Zhou et al. (2009).
327 Enhanced glyoxal concentrations have been detected over oceans in several studies (Coburn et al., 2014; Lerot et
328 al., 2010; Sinreich et al., 2010), but current models cannot reproduce this. For this reason, over oceans, we use an
329 a priori glyoxal concentration profile measured with an air-borne MAX-DOAS instrument over the Pacific Ocean
330 during the TORERO campaign (Volkamer et al., 2015).

331 3.3. Background correction

332 As already mentioned, systematic (row-dependent) biases in the retrieved SCDs often remain due to small residual
333 interferences with spectral signatures from other absorbers or due to instrumental effects. In the particular case of
334 pushbroom imaging instruments such as OMI/TROPOMI, across-track row-dependent biases (so-called stripes)
335 often occur due to the imperfect calibration of the different CCD detector rows. To reduce those biases, a
336 background correction using observations in a remote reference sector is generally applied as part of the retrieval
337 algorithm (e.g. Alvarado et al., 2014; Chan Miller et al., 2014; Lerot et al., 2010; Richter and Burrows, 2002; De
338 Smedt et al., 2018). The principle of this background correction is to add offset values to the retrieved SCDs to



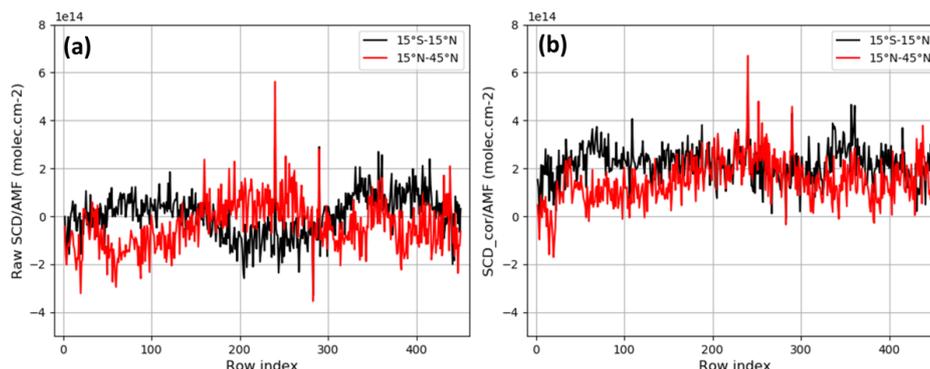
339 ensure that the resulting mean VCD in a clean remote region match an a priori known tropospheric glyoxal
340 column. Here we use the Pacific Ocean as reference sector with a constant reference VCD of 1×10^{14} molec/cm².
341 This value was chosen according to independent measurements performed in this region (Sinreich et al., 2010)
342 since current global models fail to reproduce remote sensing glyoxal levels observed over oceans (Fu et al., 2008;
343 Myriokefalitakis et al., 2008; Stavrakou et al., 2009b).

344 The background correction is applied on a daily basis in different steps:

- 345 1. First, a destriping procedure such as proposed in Boersma et al. (2007) is applied consisting in an offset
346 correction determined separately for each instrumental row, and relying on clear sky observations from
347 the Equatorial Pacific Ocean (15°S-15°N, 165°E-220°E). The offset corrections are added to all glyoxal
348 SCDs worldwide, considering their respective row.
- 349 2. Additionally to the high frequency stripes, a broadband row-dependent structure, of which the shape also
350 depends on the latitude, was identified as illustrated in Figure 4, panel (a). This figure compares the row-
351 dependence of mean uncorrected VCDs in the Pacific Ocean at Equatorial and Northern mid-latitudes.
352 The two curves are somehow anti-correlated, meaning that the destriping correction based on equatorial
353 latitudes only as applied in step 1 is not sufficient and even reinforces the mid-latitude structure. The
354 second step of the background correction aims thus at reducing this broadband row-dependent structure
355 at all latitudes while maintaining the mean latitudinal distribution of the measured background glyoxal
356 columns. For this, Pacific Ocean measurements (40°S-40°N, 165°E-220°E) are binned per 20° in latitude
357 and in groups of 15 rows in a 2-dimensional matrix. For this step, we use reference VCDs depending on
358 the latitude and resulting from the averaging of the binned VCDs along the row dimension. A
359 corresponding 2-dimensional matrix of SCD offset corrections is then computed in order, once applied
360 to the binned VCDs, the corrected values match the reference VCDs. Interpolation through this correction
361 matrix provides offsets to be applied to all SCDs retrieved worldwide.
- 362 3. Finally, the overall level of the product is adjusted with a single offset correction to ensure that the mean
363 of all clear-sky VCDs within the full reference sector (40°S-40°N, 165°E-220°E) is equal to 1×10^{14}
364 molec/cm². Panel (b) of Figure 4 shows how the identified row dependence in the VCDs at different
365 latitudes has been reduced. The general level of the columns has also been adjusted.

366

367



368

369 **Figure 4 : Row-dependence of the glyoxal vertical columns of S5p orbit #5877 (December, 1st 2018) averaged in an**
 370 **equatorial latitude band and in a Northern mid-latitude band. (a) No background correction is applied; (b) a latitude-**
 371 **dependent background correction is applied.**

372 3.4. Uncertainty estimates

373 Glyoxal tropospheric column retrievals are affected by many sources of uncertainties in the different components
 374 of the algorithm. The low glyoxal optical depth makes its retrieval highly sensitive to measurement noise and to
 375 spectral interferences with strong absorption signatures of other species or with instrumental features. Although
 376 the measurement noise can be reduced by averaging column retrievals from individual observations, spectral
 377 interferences generally lead to residual systematic errors (biases), which cannot be easily eliminated. The
 378 background correction described above aims at reducing those biases, but it has its own limitations. For example,
 379 the reference glyoxal tropospheric column within the reference sector is poorly known. In addition to spectral fit
 380 errors, there are also significant errors associated to the air mass factor calculations, mostly originating from input
 381 parameters uncertainties. For estimating the total glyoxal column error, we assume that the different error
 382 components are uncorrelated and can be summed quadratically as in (Boersma et al., 2011; Lerot et al., 2010; De
 383 Smedt et al., 2008, 2018). If the glyoxal vertical column N_v is expressed as

$$N_v = \frac{N_s - \overline{(N_{s,0} - N_{v,0,ref} \times M_0)}}{M} \quad (1)$$

384 with N_s the retrieved slant column, M the AMF, $\overline{(N_{s,0} - N_{v,0,ref} \times M_0)}$ the background correction term where $N_{s,0}$,
 385 M_0 , $N_{v,0,ref}$ are the slant columns, AMF, and the reference vertical column within the reference sector, the total
 386 glyoxal vertical column error can be written as

$$\sigma_{N_v}^2 = \frac{1}{M^2} \left(\sigma_{N_s}^2 + N_v^2 \sigma_M^2 + \sigma_{N_{s,0}}^2 + N_{v,0,ref}^2 \sigma_{M_0}^2 + M_0^2 \sigma_{N_{v,0,ref}}^2 \right) \quad (2)$$

387 where σ_{N_s} , σ_M and $\sigma_{N_{v,0,ref}}$ are the errors on the slant column, the air mass factor and the reference value used in
 388 the background correction, respectively. In the following subsections, we discuss the different contributions to
 389 each of those terms. Errors can affect the retrievals randomly or systematically (biases). While the main random
 390 error is caused by the propagation of the instrumental photon detector shot noise on the measured radiances, the
 391 other error components are considered as being systematic.



392 3.4.1. Slant column uncertainties

393 As mentioned above, the radiance measurement noise directly propagates into the glyoxal slant column retrieval
394 and leads to large random errors $\sigma_{N_s,rand}$ (or precision) due to the low glyoxal optical depth. Those are easily
395 estimated using the fit residuals RMS and the covariance matrix of the cross-sections included in the fit (Danckaert
396 et al., 2017). In the visible spectral range, the TROPOMI signal-to-noise ratio is about 1600 over dark scenes.
397 This leads to a glyoxal VCD precision (i.e. $\sigma_{N_s,rand}/AMF$) in the range of $6\text{-}10 \times 10^{14}$ molec/cm² as illustrated in
398 Figure 5, panel (d). This range of values is consistent with the scatter observed in the retrieved glyoxal SCDs in
399 regions without any significant glyoxal source. Over bright scenes, for example covered by clouds or snow, those
400 errors significantly drop because of the increased signal-to-noise ratio. For individual observations, random errors
401 dominate and averaging is needed to extract meaningful glyoxal signals.

402 There are also systematic errors associated to the DOAS spectral fit that are mainly dominated by absorption
403 cross-section uncertainties, by interferences with other species (O₄, liquid water, Ring ...), or by other effects such
404 as residual stray light. Those contributions are difficult to assess and can only be estimated from sensitivity tests
405 (Lerot et al., 2010). In general, this error term can be as high as $2\text{-}3 \times 10^{14}$ molec/cm². However, the use of a
406 radiance as reference in the DOAS fit and the application of a background correction removes a large part of the
407 systematic error in the slant column fit (see section 3). As those corrections are not always sufficient to eliminate
408 completely the SCD systematic errors due to local conditions (local pollution, residual clouds,...), we set $\sigma_{N_s,sys}$
409 to 1×10^{14} molec/cm².

410 3.4.2. AMF uncertainties

411 The errors on the air mass factor depend on the input parameter uncertainties and on the sensitivity of the air mass
412 factor to each of them. This contribution can be broken down into the squared sum (Boersma et al., 2011; Lerot
413 et al., 2010; De Smedt et al., 2018) as

$$\sigma_{M,sys}^2 = \left(\frac{\partial M}{\partial A_s} \cdot \sigma_{A_s} \right)^2 + \left(\frac{\partial M}{\partial s} \cdot \sigma_s \right)^2 + (0.15M)^2 \quad (3)$$

414 where σ_{A_s} and σ_s are typical uncertainties on the surface albedo and profile shape, respectively.

415 The contribution of each parameter to the total air mass factor error depends on the observation conditions.
416 Therefore, a small table of air mass factor derivatives spanning all observation conditions was computed using
417 VLIDORT, considering glyoxal box profile shapes with different effective heights.

418 The AMF error component related to the surface reflectivity (1st term of Eq. ((3)) is calculated using an estimated
419 uncertainty on the albedo σ_{A_s} of 0.02 (Kleipool et al., 2008). Note that this uncertainty can be occasionally larger,
420 in particular at high latitudes where snow falls may cause abrupt changes in scene albedo. The uncertainty
421 associated to the a priori profile shapes (the smoothing error) used in the retrieval is more difficult to assess,
422 especially due to the scarcity of independent glyoxal profile measurements. For every observation, an effective
423 height corresponding to the a priori glyoxal profile used in the AMF calculation is derived and used to extract the
424 appropriate AMF derivative and σ_s is taken equal to 50hPa.



425 Formulation ((3) is valid for clear sky pixels and the stringent cloud filtering we use. However, residual clouds
426 undoubtedly impact the radiative transfer and generally shield the lowermost atmospheric layers. Therefore, we
427 anticipate that the clear sky assumption generally leads to a low bias on the retrieved glyoxal columns in case of
428 residual clouds. On the other hand, the spectral interferences over bright (cloudy) scenes as discussed in section
429 3.2 impact the retrievals the other way round. The third term in equation ((3) accounts for possible errors in the
430 AMF model itself, including the neglect of aerosols and clouds, wavelength dependence, . . . , and is estimated to
431 be 15% of the air mass factor (Lorente et al., 2017).

432 3.4.3. Background correction uncertainties

433 Although the background correction is designed to overcome systematic features/deficiencies of the slant column
434 fitting, some errors are also associated to this procedure. In particular, systematic errors on the reference slant
435 columns and their air mass factors are propagated into the computed correction values. Also, there is an uncertainty
436 related to the reference glyoxal vertical column value in the reference sector. The three last terms of Eq. (2)
437 represent the total background correction uncertainty in which $\sigma_{N_{s,0}}$ is the systematic slant column error fixed to
438 1×10^{14} molec/cm² (see above section 6.5.1), and M_0 and σ_{M_0} are the air mass factors and their associated errors
439 within the reference sector. In practice, those quantities are treated similarly as the reference slant columns (i.e.
440 binned in latitude and row bins – see section 3.3). $\sigma_{N_{v,0,ref}}$ represents the error associated to the reference value
441 $N_{v,0,ref}$ and is fixed to 5×10^{13} molec/cm².

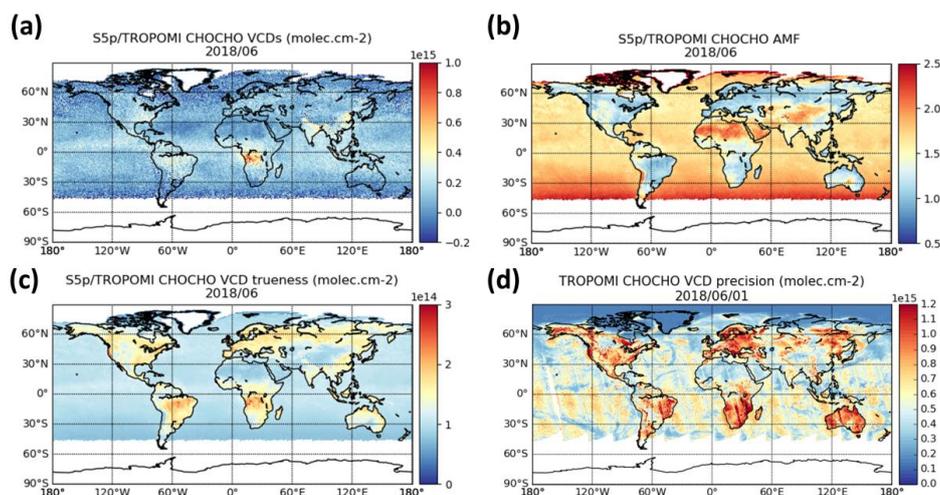
442 3.4.4. Total systematic uncertainties

443 Figure 5, panel (c) shows the estimated mean VCD systematic errors for the month of June 2018 when all
444 systematic error sources are combined together using Eq. (2). Note that the conversion of the AMF error into an
445 absolute vertical column error (2nd term of the equation) requires this error to be multiplied by the corresponding
446 vertical column. Because of the high level of noise in the product, using the retrieved column for this would lead
447 to a strong overestimation of the systematic error. To circumvent this, we use instead pre-computed climatological
448 glyoxal noise-free VCDs.

449 Total glyoxal VCD systematic errors are generally in the range $1\text{--}3 \times 10^{14}$ molec/cm², corresponding to about 30-
450 70% for emission regimes (columns larger than 2×10^{14} molec/cm²). Note that pixels strongly contaminated by
451 clouds (cloud fraction > 20%) or covered by snow/ice are discarded. Systematic errors are expected to be large
452 for those pixels mainly due to spectral interference effects (see section 3.2) and also because the information
453 content on glyoxal is reduced in case of cloud shielding. Figure 5, panel (b) shows monthly mean AMFs for the
454 same month. Small AMFs are generally caused by a priori profiles peaking near the surface, which makes the
455 retrieval more sensitive to albedo uncertainties and to a lesser extent to the a priori profile shape uncertainties.
456 This explains the anti-correlation between the AMFs and the systematic errors. In contrast, large AMFs are caused
457 either by bright surface or by background a priori profiles. For such cases, systematic errors are smaller. Note that
458 satellite column averaging kernels, defined as the Box-AMF divided by the total AMF (Eskes and Boersma, 2003),
459 are provided for every observation. They can be used to remove the smoothing error component when comparing
460 the satellite data to any other external data.



461



462

463 **Figure 5: (a) TROPOMI June 2018 monthly means of glyoxal tropospheric columns, glyoxal air mass factors (panel**
464 **(b)) and glyoxal tropospheric column systematic errors (panel (c)). Scenes contaminated by clouds or Ice/snow have**
465 **been filtered out. Panel (d) shows glyoxal tropospheric column random errors for one single day, in which all**
466 **observations have been kept to illustrate the impact of the scene brightness.**

467 4. Comparison with other satellite instruments

468 4.1. Algorithmic differences for GOME-2A/B and OMI glyoxal retrievals

469 Glyoxal tropospheric columns have also been retrieved from other satellite instruments, namely GOME-2 on
470 board the platforms Metop-A and -B and OMI on board AURA. Retrieval settings very similar to those described
471 in the previous section were applied. For GOME-2A and B, we use data records recently produced within the
472 operational environment of the EUMETSAT AC SAF (Valks et al., 2020). We list here the remaining differences
473 with respect to the TROPOMI algorithmic baseline and the specificities for each instrument.

474 All data sets essentially share the same DOAS fit settings (reference cross-sections, fit window, polynomial
475 degree...). The heterogeneity cross-sections are omitted for the GOME-2 and OMI retrievals. While the
476 instrumental design of GOME-2 makes it weakly sensitive to scene heterogeneity, it would be beneficial for OMI
477 to include similar cross-sections but that would imply a reprocessing of the complete slant column data set data
478 with limited added-value for the large-scale comparison with TROPOMI that we present in the next subsection.
479 For the GOME-2 instruments, we also fit two additional cross-sections representative of the instrumental
480 sensitivity to light polarization as provided from the level-1 key data (EUMETSAT, 2011) as well as one pseudo
481 cross-section to account for an along-track spectral resolution change occurring due to instrumental temperature
482 change (Azam and Richter, 2015). Note that for GOME-2 the cross-sections are convolved with an instrumental
483 slit function optimized as part of the wavelength calibration for every measured irradiance (De Smedt et al., 2015),
484 which allows accounting for the known long-term drift of the GOME-2 instrument spectral response function.



485 Differences in air mass factor calculations consist only in using, over land, a priori profiles provided by
486 IMAGESv2, the chemical transport model predecessor of MAGRITTE, at the coarser resolution of $2.0^{\circ} \times 2.5^{\circ}$. For
487 the GOME-2 instruments, we use the directionally dependent Lambertian-equivalent reflectivity database
488 produced by Tilstra et al. (2021) instead of the OMI database.

489 A background correction procedure is applied consistently with the one used for TROPOMI. The GOME-2
490 instruments being whiskbroom scanners, there is no destriping procedure as such but instead a viewing zenith
491 angle-dependent correction is applied, also relying on the slant columns retrieved in the Equatorial Pacific sector.
492 This correction may account for example for remaining biases related to the instrumental polarization sensitivity.
493 For both OMI and GOME-2, the row/VZA dependence does not show any obvious change along the orbit and the
494 corresponding correction thus relies only on the low latitude measurements.

495 Note that the OMI and GOME-2 glyoxal products are filtered for cloudy scenes using cloud fraction lower than
496 20% as taken from the O_2-O_2 (Veeffkind et al., 2016) and OCRA (Lutz et al., 2016) cloud products, respectively.
497 The empirical correction for strong NO_2 absorption signature described in section 3.1 has been applied to those
498 instruments as well. In the following section, we compare the TROPOMI glyoxal retrievals with the OMI and
499 GOME-2A/B data sets. The OMI record covers the period 2005-2018, while GOME-2A/B span the periods 2007-
500 2017 and 2013-2020, respectively. OMI and GOME-2A records were interrupted when their respective quality
501 was degraded too severely and other instruments were available to continue the morning and afternoon time series.

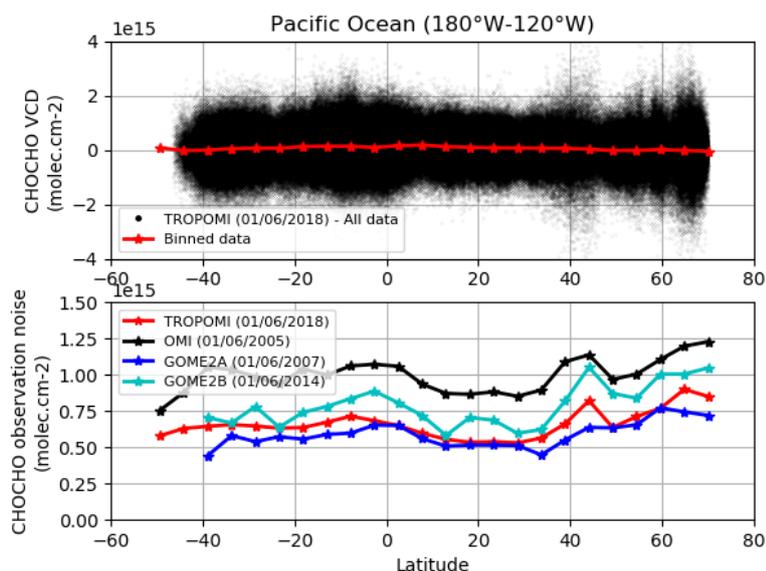
502 **4.2. Glyoxal satellite inter-comparison**

503 **4.2.1. Comparison of the noise level**

504 As mentioned before, the level of noise in the satellite glyoxal tropospheric column products is large compared to
505 the real signal. This is illustrated in the upper panel of Figure 6 which shows all individual clear-sky TROPOMI
506 glyoxal columns retrieved in the Pacific Ocean on June, 1st 2018 and plotted as a function of their latitude. The
507 scatter is significant ($\sigma \approx 5-7 \times 10^{14}$ molec/cm²) with respect to the small glyoxal VCDs averaged in 5° -latitude
508 bins in this sector. The lower panel compares the standard deviation of the retrievals from TROPOMI, OMI,
509 GOME-2A and B in the same remote sector for the 1st of June of their respective first year of operation. The
510 scatter in the retrievals is directly linked to the instrumental signal-to-noise ratio, which is documented to be
511 around 500 for OMI (Schenkeveld et al., 2017), 1000 for GOME-2 (Zara et al., 2018) and 1500 for TROPOMI
512 (Kleipool et al., 2018). In practice, we see that the CHOCHO observation noise is indeed slightly larger for OMI,
513 that GOME-2B retrievals are noisier than those from GOME-2A, which have a level of noise similar to
514 TROPOMI. Considering the very small footprint size of TROPOMI (3×7.5 km² and 3×5.5 km² after August 2019)
515 compared to the other instruments (GOME-2: 80×40 km²; OMI: 13×24 km² at nadir), the TROPOMI observation
516 noise is remarkably low. More importantly, the much larger amount of TROPOMI data compared to OMI (~15x)
517 and GOME-2 (~100x) allows maintaining a better time or spatial resolution for a given target noise level. For
518 example, the random error associated to the daily glyoxal column averaged in an area defined by a circle with a
519 radius of 50 km will be less than 0.5×10^{14} molec.cm⁻² for TROPOMI, while it will remain larger than 2.5×10^{14}
520 molec/cm² and 4.0×10^{14} molec/cm² for OMI and GOME-2, respectively.



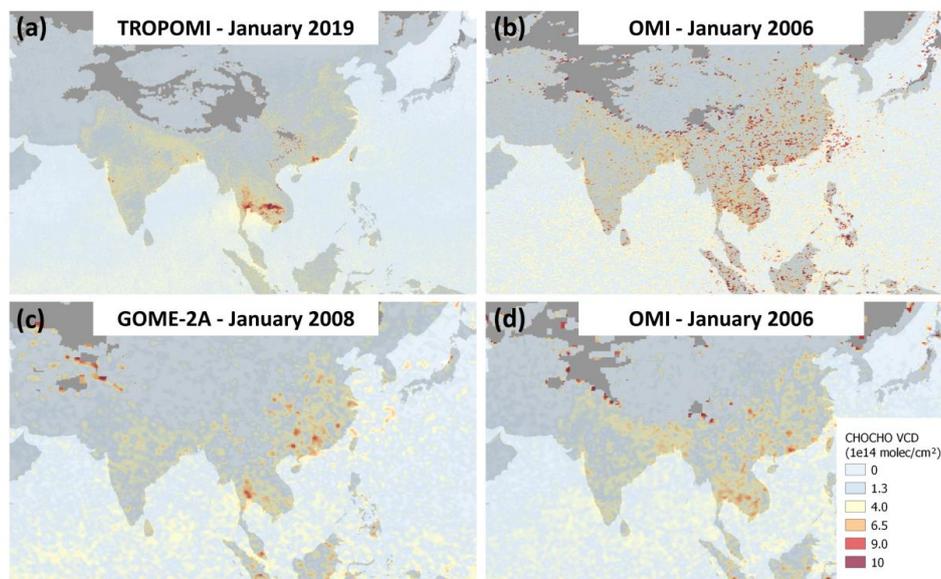
521 This is illustrated in Figure 7, which compares January monthly mean glyoxal VCD fields over Asia at the
522 resolution of 0.05° for TROPOMI and OMI (upper panels) and 0.25° for GOME-2A and OMI (lower panels) after
523 one year of their respective operation. At the resolution of 0.05° , the level of noise in the TROPOMI glyoxal map
524 is very low and many details can be distinguished in the glyoxal spatial distribution. In particular, hot spots of
525 glyoxal over many megacities are clearly identified (e.g. over Bangkok, New Delhi, Ho Chi Minh City,
526 Shenzhen...) but also over Cambodia where large fires occur every year from January to March. At this resolution
527 of 0.05° , the level of noise in the OMI map remains high and prevents distinguishing such details. At the coarser
528 spatial resolution of 0.25° , the reduction of the noise in the OMI and GOME-2 monthly glyoxal fields appears to
529 be sufficient to better distinguish the glyoxal spatial distribution but at the cost of a significant smoothing. In the
530 next section, we will intercompare the four satellite products at low temporal and spatial resolution in order to
531 minimize the impact of the noise and to identify possible systematic discrepancies.



532

533 **Figure 6** : Illustration of the level of noise in satellite CHOCHO VCD retrievals. The upper panel shows all clear sky
534 individual glyoxal VCDs retrieved in the Pacific Ocean from TROPOMI observations on June, 1st 2018. The scatter is
535 very large compared to the low real signal as illustrated by the data binned in 5° latitude bands. The lower panel
536 compares the standard deviation of the retrievals in the same sector from TROPOMI, OMI, and GOME2A/B on the
537 1st of June of their respective first year of operation.

538



539

540 **Figure 7: Illustration of the impact of the instrumental signal-to-noise and available amount of data on monthly mean**
541 **glyoxal VCD fields retrieved from different satellite instruments. The (a) TROPOMI data for January 2019 gridded**
542 **at a resolution of 0.05°, (b) OMI data for January 2006 gridded at a resolution of 0.05°, (c) GOME-2A data for**
543 **January 2008 gridded at a resolution of 0.25°, and (d) OMI data for January 2006 gridded at a resolution of 0.25°.**
544 **Cloudy scenes have been filtered out and a smoothing filter has been applied on the four presented fields based on a**
545 **spatial mean with the nearest neighbouring grid cells.**

546

4.2.2. Comparison of mean glyoxal fields

547

548 First, Figure 8 and Figure 9 compare seasonal maps of glyoxal VCDs generated from TROPOMI, OMI and
549 GOME-2A/B data products. In order to reduce the data scatter for each instrument, those maps are based on long
550 time series as indicated in the figures. Therefore, a one-to-one match is not expected. As can be seen, the
551 consistency between the four instruments is excellent. Glyoxal patterns are captured similarly for all seasons in
552 terms of both spatial distribution and VCD values. The largest glyoxal columns are observed in tropical regions,
553 where biogenic emissions are important, and in regions with important fire events (e.g. Amazonia and Northern
554 Africa in SON, Thailand/Indochina in MAM, Western US in August,...). At mid-latitudes, the glyoxal columns
555 follow the seasonal cycle of biogenic activity with maximum values during summertime. Localized hot spots of
556 glyoxal are visible over megacities corresponding to strong anthropogenic emissions (e.g. Northern China Plain,
557 Bangkok, Teheran, New Delhi, Sao Paulo...). Also a persistent oceanic glyoxal signal is seen consistently by the
558 four sensors. Note that a similar signal has been detected from ship- and airborne MAX-DOAS in the equatorial
559 Pacific and Atlantic Oceans (Behrens et al., 2019; Sinreich et al., 2010; Volkamer et al., 2015). In contrast to
560 TROPOMI and OMI, the level of noise in the GOME-2 data sets significantly increases over the South Atlantic
561 Anomaly despite the application of a spike-removal procedure (section 3.1). Overall the GOME-2B maps are
562 noisier than those from other sensors due to the lower signal-to-noise ratio of the spectra and a shorter time series.
Compared to the UV, the sensitivity to the surface is larger in the visible, which may introduce interferences with



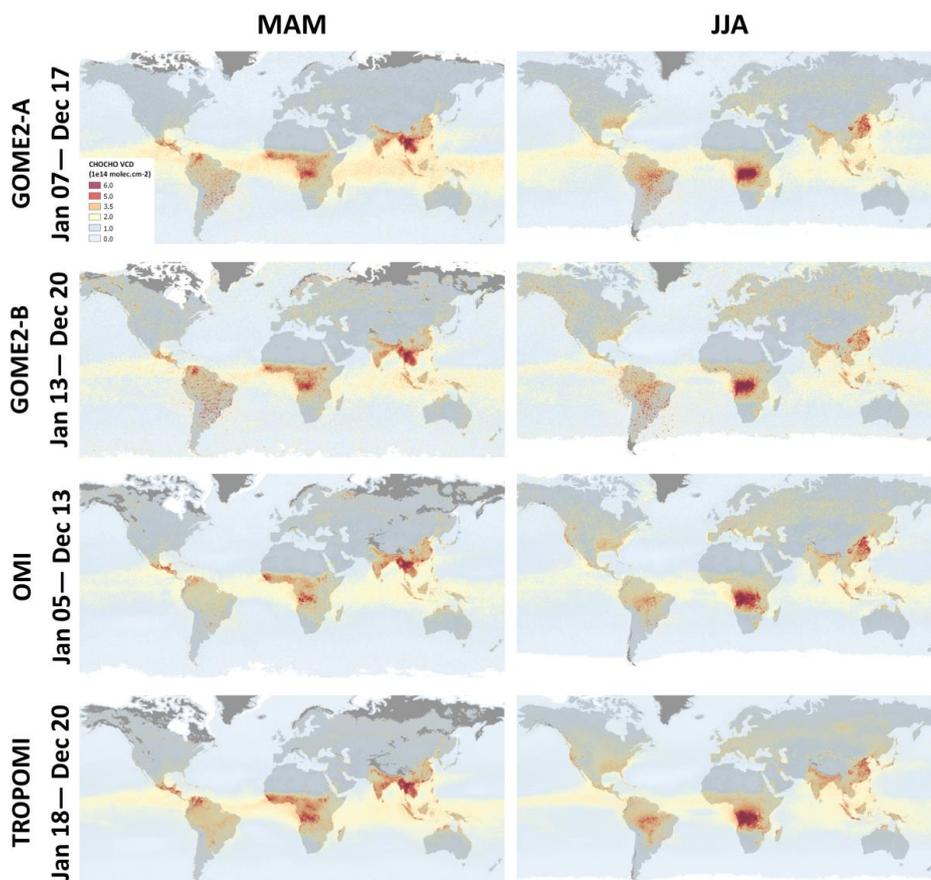
563 the spectral signature of specific ground surfaces, and thus may potentially lead to a bias on the retrieved columns.
564 A striking example is over the Kara-Bogaz-Gol near the Caspian sea, which is one of the saltiest lakes in the world
565 and contains large concentrations of sediments (Kosarev et al., 2009). The glyoxal signal detected over that lagoon
566 is unlikely to be physical and likely originates from interferences with the ground reflectance spectral signature.

567

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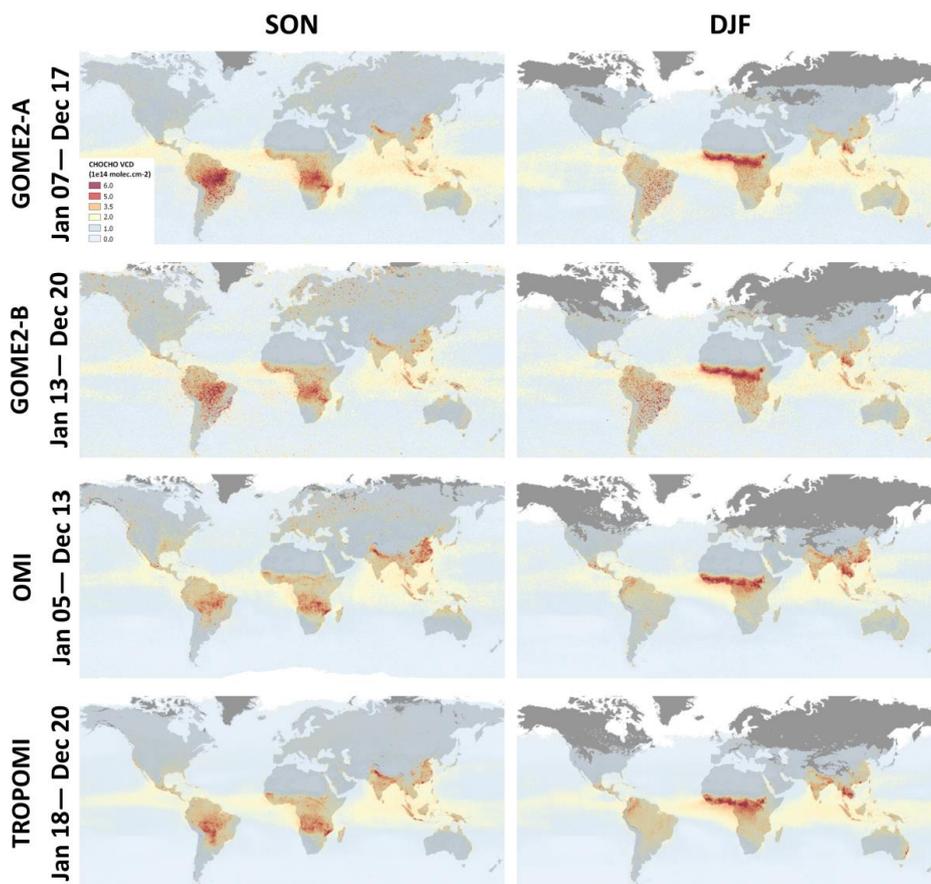
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571

572 **Figure 8: Comparison of long-term averaged global CHOCHO VCDs (in 10^{14} molec/cm²) derived from**
573 **GOME-2A, GOME-2B, OMI and TROPOMI sensors, for the March-April-May period (left panels) and**
574 **the June-July-August period (right panels).**



575

576 **Figure 9: Comparison of long-term averaged global CHOCHO VCDs (in 10^{14} molec/cm²) derived from**
577 **GOME-2A, GOME-2B, OMI and TROPOMI sensors, for the September-October-November period (left**
578 **panels) and the December-January-February period (right panels).**

579 For a more detailed investigation of the consistency of the TROPOMI data set with OMI and GOME-2A/B, we
580 compare complete time-series of monthly median glyoxal columns in selected regions (shown in Figure 10).

581 The red rectangles indicate the regions on which we focus in Figure 11 and Figure 12, while the global statistics
582 for all highlighted regions are given in Figure 13. Detailed figures are provided for all regions as supplementary
583 material (Figures S1, S2, S3, S4). Figure 11 compares directly the four full time series, while Figure 12
584 compares the typical climatological seasonal variations as obtained by combining all available years. The error
585 bars in the latter figure represent the interannual variability, and the 2-sigma standard deviation of the four
586 satellite products is indicated as inset.

587 In the Tropics (e.g. Amazonia, Equatorial/North Central Africa), the four data sets are relatively stable over time.
588 All instruments observe similar seasonal cycles and column values, although OMI appears to be slightly lower

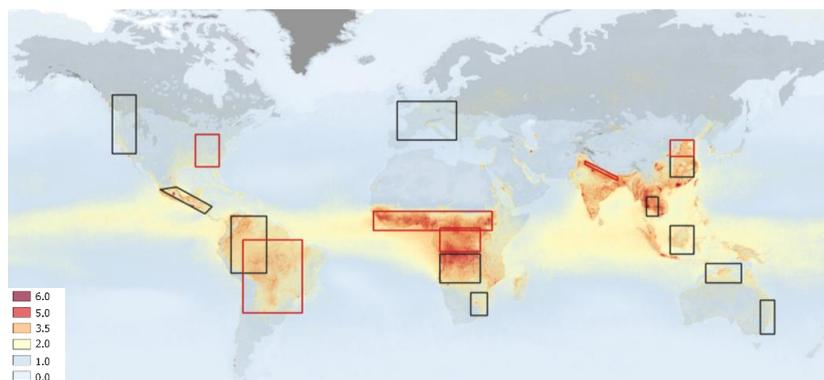


589 than the others, in particular in Equatorial Africa. The inter-annual variability in Amazonia is high compared to
590 other regions worldwide. Glyoxal is produced in that region to a large extent by fire emissions, which are highly
591 variable. There is a direct correlation between years with high glyoxal columns and large fire emissions (e.g. 2007,
592 2010, 2015, 2019) as derived from the GFED database (van Der Werf et al., 2017;
593 <https://www.globalfiredata.org/>). Interestingly, glyoxal columns measured by the morning GOME-2 instruments
594 are larger than the OMI columns in the early afternoon during the fire seasons. This is consistent with the diurnal
595 variation measured in satellite HCHO columns by De Smedt et al. (2015) and would deserve further investigation.
596 Other regions display a more regular seasonal cycle, consistently seen by the four instruments.

597 In Asia, there are many hot spots, of which the origin is manifolds and strongly depends on the region and season.
598 In addition to biogenic activities, large emissions due to fires may significantly contribute to the glyoxal columns.
599 As illustrated in Figure 12, in the Indo-Gangetic Plain, there are typically two fire seasons in April/May and in
600 October/November (after the Monsoon period) related to agricultural burning of wheat residue (Kumar et al.,
601 2016), and leading to two maxima in the glyoxal VCD seasonal cycle with a significant interannual variability.
602 For example, during the COVID-19 Indian lockdown in April/May 2020, fire activity has been reduced leading
603 to smaller emissions (Levelt et al., 2021). This region is also highly populated, causing large emissions due to
604 human activities. This is also true in North-East China where glyoxal columns remain significant in winter, while
605 biogenic emissions are low during that period of the year. Although less variable than fire emissions,
606 anthropogenic emissions may also change over time. Despite those variable emissions, the four data sets spanning
607 different time periods show a high level of consistency. In China, it seems that the glyoxal columns as observed
608 by OMI, GOME-2A and B are slightly reduced after 2014. This would deserve further investigation. On the other
609 hand, any interpretation based on long-series of OMI data must be treated carefully since the instrument suffers
610 from an evolving row anomaly (Schenkeveld et al., 2017), which impacts the stability of the product and causes
611 an increasing number of outliers, especially at mid-latitudes. For example, over the Indo-Gangetic Plain, the OMI
612 columns deviate regularly from the other instruments after 2014. In general, remnants of noise are also visible in
613 the GOME-2 time series, which show somewhat less smooth time series than TROPOMI.

614 At mid-latitudes, the lower sun elevation, especially during local wintertime, makes the retrievals more
615 challenging. Nevertheless, a small maximum is consistently observed during the local summertime. During
616 wintertime, TROPOMI columns appear slightly lower than those from the other satellites. As mentioned before,
617 the stronger impact of the row anomaly at mid/high-latitudes leads to a larger number of outliers in the OMI data
618 set and to a low bias in winter after 2013/2014.

619

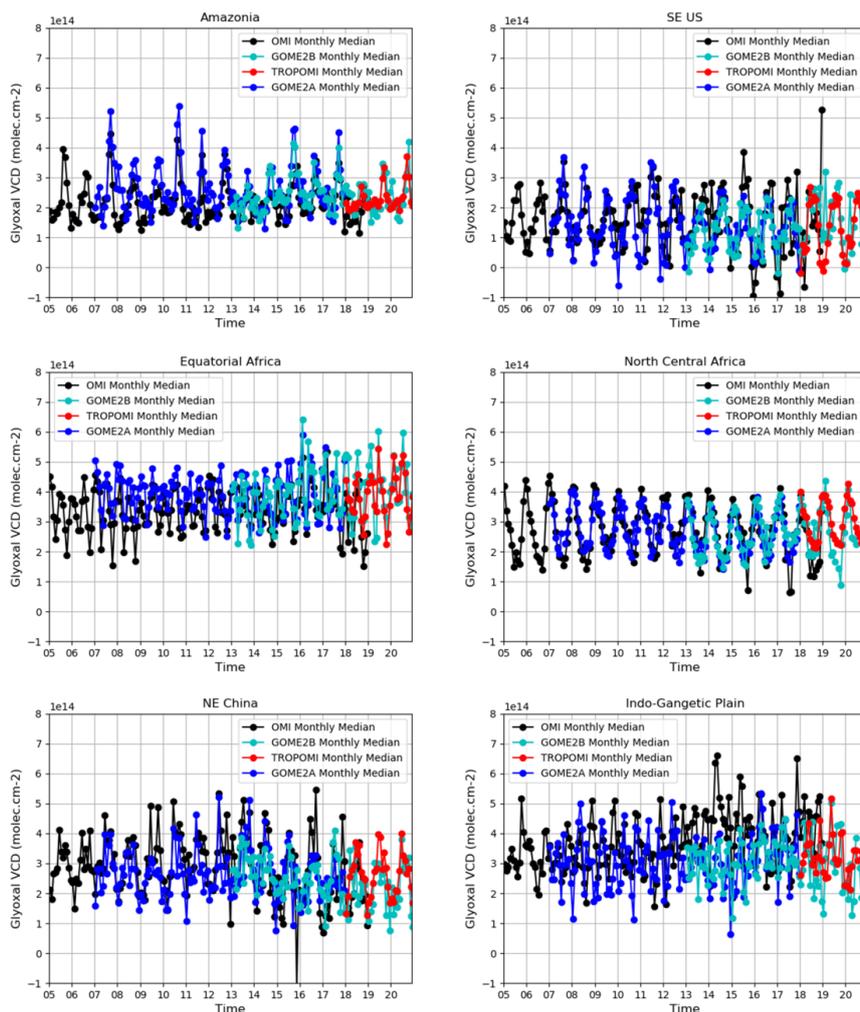


620

621 **Figure 10 : TROPOMI glyoxal VCD distribution (in $1e14$ molec/cm²) averaged on the period January 2018-December**
622 **2020. The rectangles represent the regions where the glyoxal products from different satellites are intercompared with**
623 **a specific focus on red regions in Figure 11 and Figure 12.**

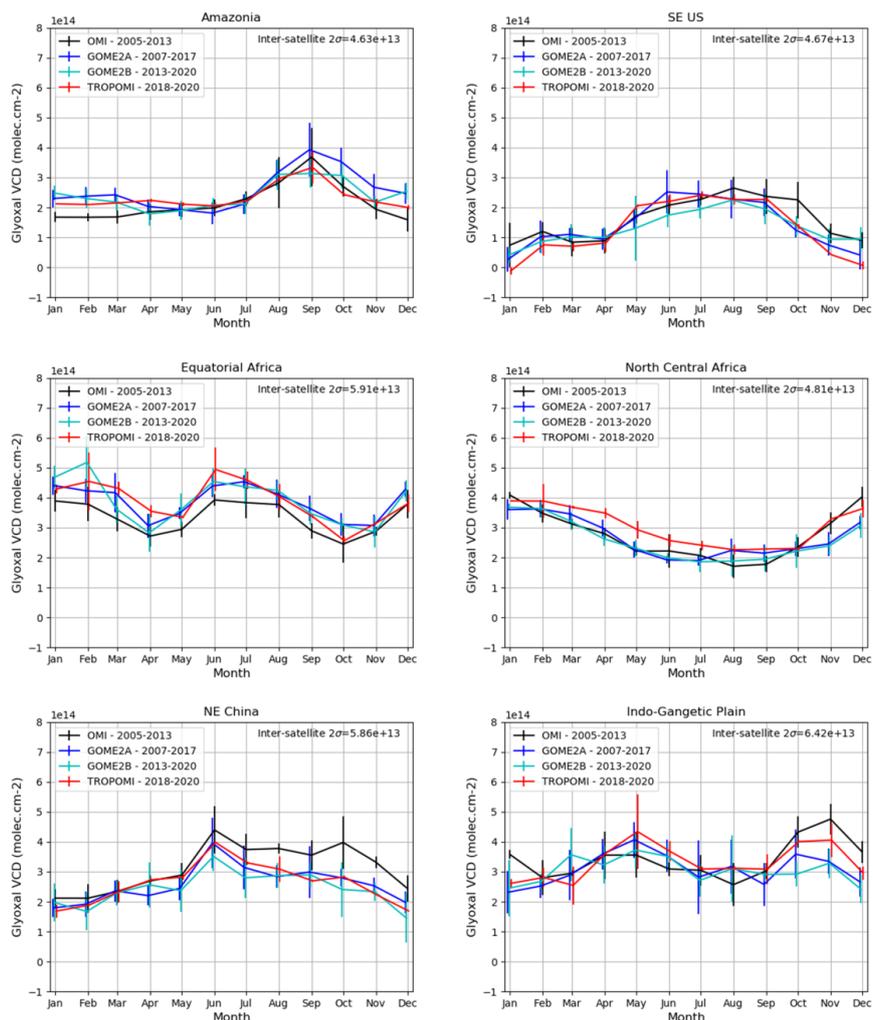
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625



626

627 **Figure 11: Comparison of the monthly median glyoxal VCD time series from GOME-2A/B, OMI and TROPOMI in a**
628 **few selected regions worldwide.**



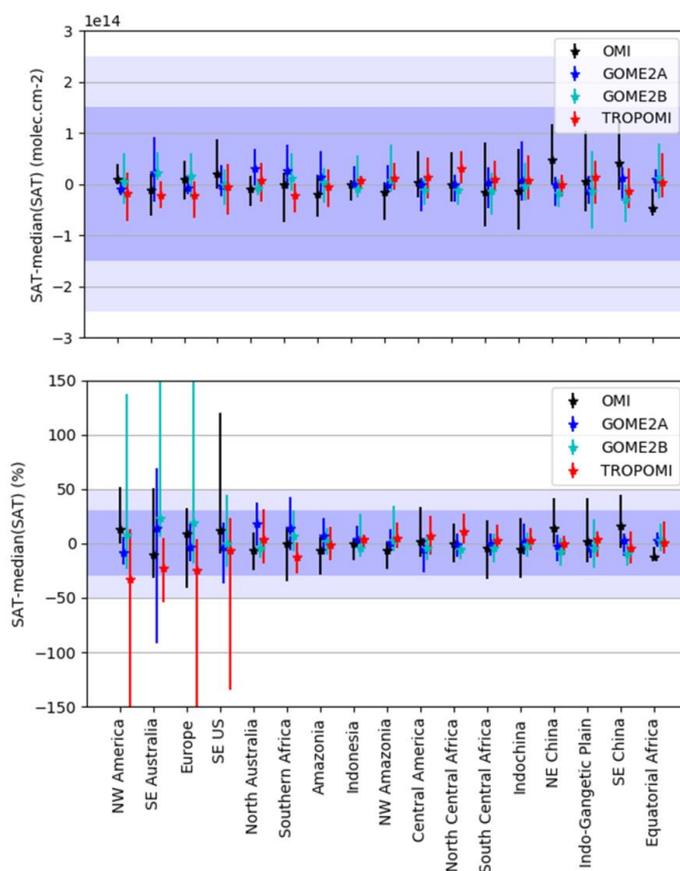
629

630 **Figure 12: Comparison of the climatological seasonal variation of the monthly median glyoxal VCDs from GOME-**
631 **2A/B, OMI and TROPOMI in a few selected regions worldwide. The error bars represent the interannual variability**
632 **as derived from the full time series.**

633 Figure 13 summarizes for all regions drawn in Figure 10 the absolute and relative deviation of each of the four data
634 sets with respect to the median values of the ensemble. The symbols represent the median deviation considering
635 all months of the year, while the error bars represent the full range of the monthly deviations. Regions are sorted
636 by increasing mean glyoxal vertical column amounts and light and dark blue shaded areas indicate 2.5×10^{14}
637 molec/cm^2 (50%) and 1.5×10^{14} molec/cm^2 (30%) differences as guidelines. Inter-satellite deviations are generally
638 less than 5×10^{13} molec/cm^2 (20%). The large error bars in the relative differences plot for mid-latitude regions are
639 caused by local wintertime months during which the glyoxal content is very low, if not negligible, and are therefore
640 meaningless. Overall, the inter-satellite consistency of the glyoxal VCD products is excellent. In the next section,



641 we will investigate the product quality with comparisons with independent ground-based MAX-DOAS glyoxal
642 observations at a few stations in Asia and Europe.



643
644 **Figure 13: Median deviation of the glyoxal VCD differences for TROPOMI, OMI, GOME-2A/B against the median**
645 **value of the ensemble of the four data sets in the selected regions worldwide drawn in Figure 10. Those are plotted in**
646 **absolute values (molec/cm²) in the upper panel and in relative values (%) in the lower panel. The error bars indicate**
647 **the full range of the deviations considering climatological monthly data. Regions are sorted by increasing median**
648 **glyoxal VCD value from left to right. The light and dark blue shaded area indicate differences of 1.5 molec/cm² (30%)**
649 **and 2.5 molec/cm² (50%).**

650

651 5. Validation with MAX-DOAS data

652 5.1. Description of MAX-DOAS data sets and methodology

653

654 MAX-DOAS instruments measure scattered solar light in the UV-Visible spectral range at different elevation
655 angles above the horizon and allow retrieving information on trace gases and aerosol extinction in the altitude



656 range below 2-3km of the atmosphere, where the instrumental sensitivity is the highest. In a first approximation,
 657 vertical columns of boundary layer gases can be estimated from MAX-DOAS measurements using a simple
 658 geometrical approach (Brinksma et al., 2008; Hönninger et al., 2004). More elaborated approaches exploit a set
 659 of different elevation angles to derive information on the vertical distribution of the gas concentration with up to
 660 4 degrees of freedom, resulting in more accurate vertical columns in the 0-4 km altitude range (e.g. Beirle et al.,
 661 2019; Clémer et al., 2010; Irie et al., 2011; Friedrich et al., 2019).

662 Glyoxal concentrations can be derived from MAX-DOAS measurements in the visible range. However, the
 663 number of glyoxal MAX-DOAS data sets is very limited, especially those covering a period long enough to allow
 664 the validation of satellite data during entire seasonal cycles. Moreover, MAX-DOAS retrievals are affected by
 665 similar difficulties as satellite retrievals (noise, spectral interferences). Here, we collected an ensemble of data
 666 sets from nine stations located in Asia and Europe (see Table 3) spanning at least one year. Altogether a wide
 667 range of glyoxal columns and emission regimes are covered by those stations. Unfortunately, the approach to
 668 retrieve glyoxal from MAX-DOAS has not been homogenized so far, and they cannot be considered as true
 669 fiducial reference measurements. For example, although the same interfering species have been included in the
 670 DOAS fits, the reference cross-section data as well as the fitting interval may vary. The design (spectral range,
 671 spectral resolution, detector type, etc.) and operation mode of the instruments differ substantially, resulting in
 672 different sensitivities to changes in retrieval settings. Finally, the slant-to-vertical column conversion is performed
 673 differently from one station to another (see Table 3). Despite those limitations, the comparison of glyoxal
 674 tropospheric columns from satellites with nine different MAX-DOAS instruments is unprecedented.

675 Among the available MAX-DOAS data sets, three (Xianghe/China, Chiba/Japan and Phimai/Thailand) are long
 676 enough to allow a comparison with OMI and GOME-2A/B in addition to TROPOMI. The other ones span shorter,
 677 and more recent periods and will be used only for comparison with the TROPOMI product. The Xianghe station
 678 has the longest and stable data record, and provide vertical profiles of glyoxal. Therefore we have used this
 679 reference station to perform a thorough analysis of the satellite product stability and of the impact of applying
 680 satellite averaging kernels. For the other stations, we performed a more qualitative comparison of the seasonal
 681 cycles of the glyoxal tropospheric columns. For the data collocation, we select MAX-DOAS data ± 1.5 hour around
 682 the satellite overpass time and satellite data within a radius of 100 km (150 km for Phimai) and 20 km around the
 683 station for GOME-2A/B/OMI and TROPOMI, respectively. Daily median glyoxal columns are computed if both
 684 satellite and ground-based data are available and finally monthly medians of the daily median columns are
 685 compared.

686 **Table 3 : List of MAX-DOAS stations used in the study and brief description of the approach to generate the glyoxal**
 687 **data.**

Station (coordinates) Time range	Institution PI	Retrieval Approach and fit interval	Reference
Xianghe/China (39.75°, 116.96°E) 2010-2020 Uccle/Belgium (50.78°N, 4.35°E) 2017-2020	BIRA-IASB	Profile retrieved using an Optimal Estimation scheme 436-468 nm	(Clémer et al., 2010; Hendrick et al., 2014)
Chiba/Japan (35.63°N, 140.10°E) 2012-2020	CERES	Profile retrieved using a parametrization approach	(Hoque et al., 2018; Irie et al., 2011)



Phimai/Thailand (15.18°N, 140.10°E) 2014-2020 Pantnagar/India (29.03°N, 79.47°E) 2017-2020		436–457 nm	
Mohali/India (30.67°N, 76.73°E) May 2019 - 2020	MPIC/IISERM	Profile retrieved using a parametrization approach 400-460 nm	(Beirle et al., 2019; Kumar et al., 2020)
Athens/Greece (38.05°N, 23.80°E) 2018-2020 Vienna/Austria (48.18°N, 16.39°E) 2018-2020 Bremen/Germany (53.11°N, 8.86°E) 2018-2020	IUP-UB	Columns retrieved using the Geometrical Approximation 436-468 nm	(Alvarado et al., 2020b; Gratsea et al., 2016; Schreier et al., 2020)

688

689

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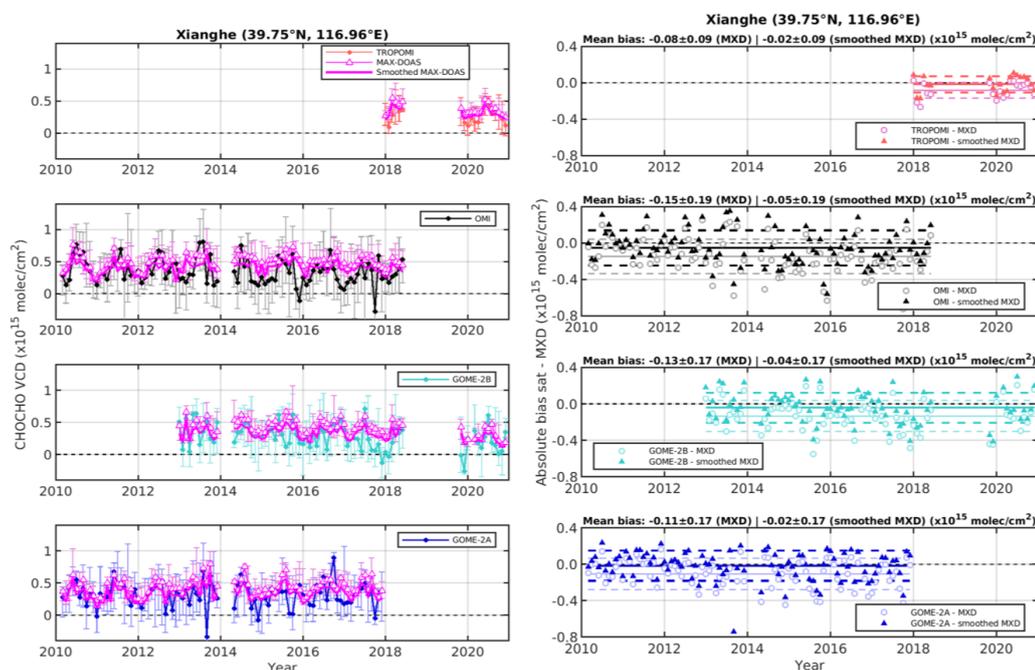
5.2. Validation results

691

692 Figure 14 focuses on the comparison of monthly median glyoxal tropospheric columns retrieved from TROPOMI,
 693 OMI, GOME-2A and GOME-2B with columns from the BIRA-IASB MAX-DOAS instrument in Xianghe
 694 (China). The left panels compare the full time series for each satellite sensor with the MAX-DOAS data record.
 695 The right panels show the corresponding satellite/MAX-DOAS absolute differences. Note that the MAX-DOAS
 696 measurements have been interrupted from mid-2018 to mid-2019 due to an instrumental problem. Overall, all four
 697 satellite instruments reproduce quite well the seasonal cycle seen by the MAX-DOAS instrument. However for
 698 all of them, except for the recent TROPOMI, a degradation appears after a few years of operation. For OMI, the
 699 consistency with the MAX-DOAS is excellent before 2013, but the number of outliers increases afterwards and
 700 the columns during wintertime become too low. This is attributed to the evolving row anomaly as discussed in
 701 section 4.2. The GOME-2A/B data sets also agree quite well with the ground-based data for their first years of
 702 operation but then suffer from an increasing number of outliers after 2014 and 2017, respectively. Nonetheless,
 703 the quality of the data sets remains very reasonable. The consistency of the TROPOMI time series with the MAX-
 704 DOAS is also excellent and is characterized by a smooth temporal variability without any outliers on a monthly
 705 basis. The absolute differences shown in the right panels also clearly indicates a reduced scatter compared to the
 706 other satellites, despite the fact that a smaller overpass radius of 20 km was used instead of 100 km. This is
 707 reflected in the standard deviation of the differences given in the titles of each subpanels. The TROPOMI standard
 708 deviation is 0.9×10^{14} molec/cm², while it is larger than 1.7×10^{14} molec/cm² for other sensors. On average, there
 709 are small negative biases with respect to the MAX-DOAS data for the four satellite time series (also given in the
 710 panel titles), ranging between -0.8×10^{14} molec/cm² for TROPOMI and -1.5×10^{14} molec/cm² for OMI. For this
 711 particular station, we investigated the impact of applying the satellite averaging kernels to smooth the MAX-
 712 DOAS glyoxal profiles. This process allows simulating MAX-DOAS columns which would be retrieved from the
 713 satellite algorithm, considering its own a priori profile information. The comparison of the satellite columns with
 714 the smoothed MAX-DOAS data therefore removes differences due to imperfect satellite a priori profile



715 information. As shown in Figure 14, smoothing the MAX-DOAS columns reduces the satellite/MAX-DOAS bias
 716 to values ranging from -0.2×10^{14} molec/cm² (TROPOMI) to -0.5×10^{14} molec/cm² (OMI).

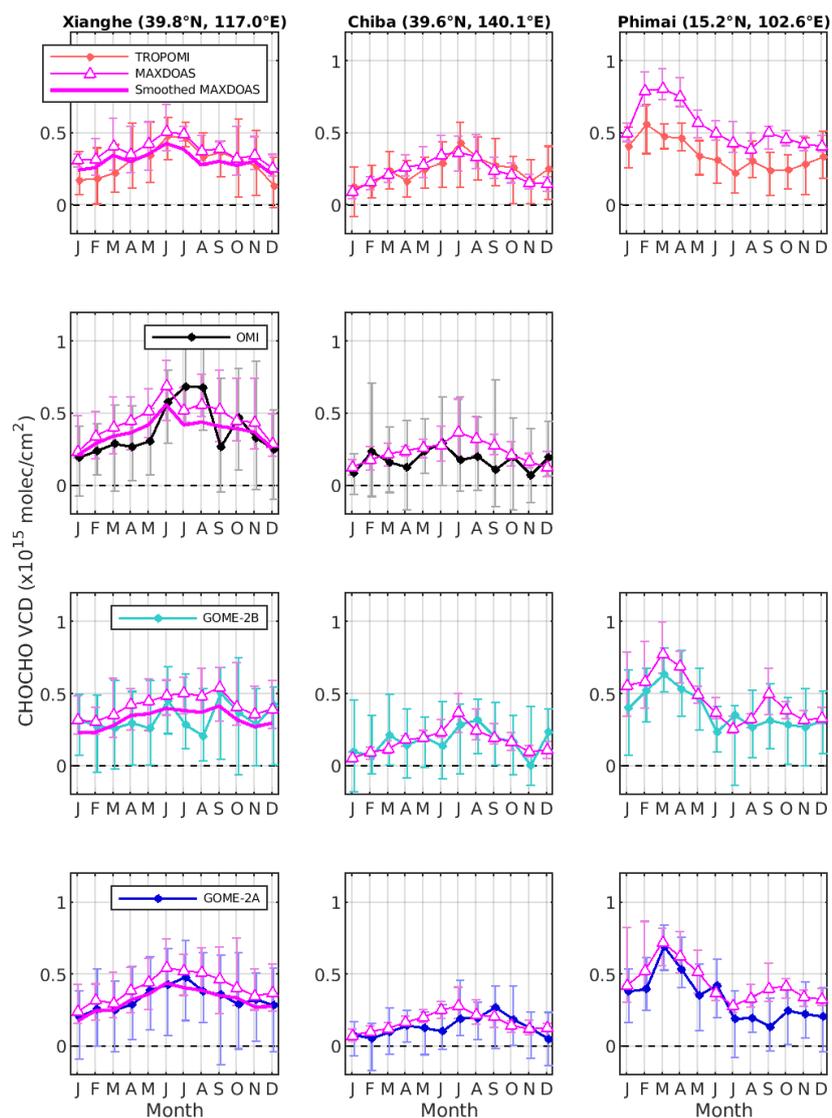


717
 718 **Figure 14 : Comparison of the monthly median glyoxal tropospheric vertical columns retrieved from satellite and**
 719 **MAX-DOAS (MXD) instruments in Xianghe (China). The four left panels compare the time series from TROPOMI,**
 720 **OMI and GOME-2A/B with the MXD time series. MXD columns are also shown when smoothed with the satellite**
 721 **averaging kernels. The error bars represent the 25 and 75% percentiles. The four right panels show the corresponding**
 722 **time series of the satellite-MD absolute differences. Both original and smoothed MXD data are shown. Mean bias and**
 723 **standard deviation of the differences are given in the panel titles and are also represented in the right panels with the**
 724 **full and dashed coloured lines.**

725
 726 In Figure 15, we compare the median satellite and MAX-DOAS seasonal cycles of the glyoxal tropospheric
 727 columns at three stations (Xianghe, Chiba and Phimai) where the time series present a good overlap with the OMI
 728 and GOME-2A and B records, in addition to TROPOMI. In Xianghe, the seasonal cycle of the smoothed MAX-
 729 DOAS columns is also shown, illustrating again the reduction of the satellite/MAX-DOAS bias when the a priori
 730 profile error component is removed. Note that the OMI and GOME-2B seasonal cycles are computed using data
 731 until end of 2013 and 2016 to limit the impact of the increasing number of outliers. In each comparison panel, the
 732 MAX-DOAS cycle is always computed using the same time range as the satellite instrument. Overall, the seasonal
 733 patterns are consistently captured by the satellite and MAX-DOAS instruments. In Xianghe, the GOME-2A and



734 TROPOMI cycles follow closely the MAX-DOAS curves, although TROPOMI slightly underestimates the MAX-
735 DOAS columns during winter months. OMI and GOME-2B also reproduce the general seasonal pattern but show
736 a somewhat more scattered curve, likely due to their slightly less stable time series. In Chiba where the glyoxal
737 signal is mostly driven by biogenic emissions, the agreement between the satellites and the MAX-DOAS
738 measurements is excellent both in terms of variability and absolute values. Again, OMI shows a larger scatter (as
739 also indicated by the larger error bars representing the inter-annual variability). In Phimai, where pyrogenic
740 emissions are responsible for large glyoxal columns especially in the first few months of the year, the seasonal
741 variability seen by the satellites and the MAX-DOAS is very consistent. A negative bias larger than for other
742 stations is nevertheless observed. This can be related to other studies that identified larger biases in NO₂ or HCHO
743 DOAS products for elevated column conditions (e.g. De Smedt et al., 2021; Verhoelst et al., 2021; Vigouroux et
744 al., 2020). Possible causes for such biases are the different air masses probed by the satellite and ground-
745 based instruments, their different vertical sensitivity as well as the a priori vertical profile information
746 used in the retrieval algorithms.



747

748 **Figure 15 :** Comparison of the monthly median glyoxal tropospheric vertical column seasonal cycle as retrieved from
 749 **TROPOMI, OMI, GOME-2A/B and MXD** in Xianghe (China), Chiba (Japan) and Phimai (Thailand). The columns
 750 correspond to the three stations and the rows to the different satellites. In Xianghe, MXD data smoothed with the
 751 satellite averaging kernels are also shown. The error bars represent the interannual variability (25% and 75%
 752 percentiles based on the full time series available). Note that the comparison of with the MAX-DOAS data in Phimai is
 753 not shown as the latter starts in 2014 when OMI is degraded.

754

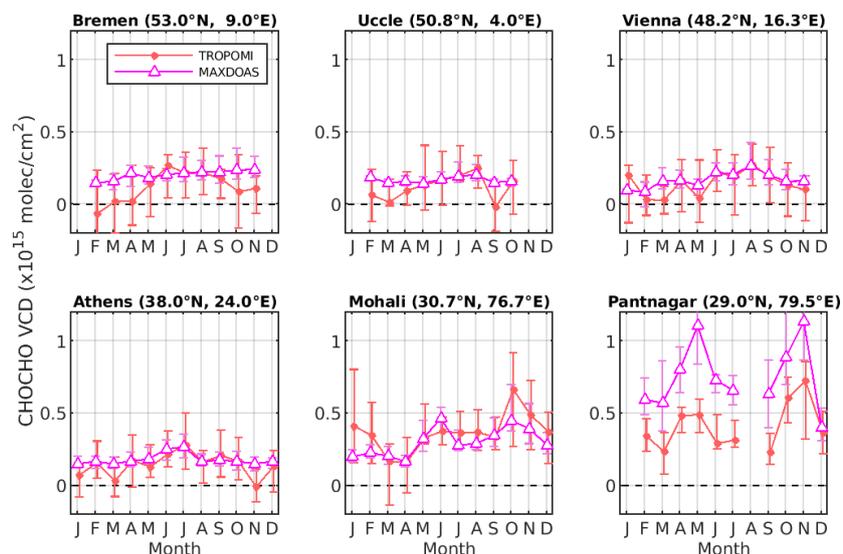


755 In Figure 16, we compare again the seasonal cycle of glyoxal VCDs retrieved from TROPOMI with that from
756 more recent MAX-DOAS time series at six different stations. Four of them are located at mid-latitude in Europe
757 and show relatively low glyoxal columns, while larger average values are measured at the two other stations, in
758 Northern India (Mohali and Pantnagar). In Vienna/Austria and Athens/Greece, TROPOMI and MAX-DOAS
759 glyoxal columns agree very well and show consistent seasonal dependencies with maximum and minimum values
760 during summertime and wintertime, respectively. On the other hand, at the higher latitude stations of
761 Bremen/Germany and Uccle/Belgium, the consistency of the seasonal variations seen from space and from the
762 ground is somewhat poorer. While the glyoxal columns agree well during summertime, the satellite columns tend
763 to underestimate MAX-DOAS values in winter, the latter showing almost no seasonal variation. Satellite glyoxal
764 retrievals at those latitudes are challenging in winter because of the low sun elevation causing a reduced sensitivity
765 to the lowermost atmospheric layers. As mentioned in section 3.2, observations with solar zenith angles larger
766 than 70° are filtered for this reason, which explains the gap between November and January at those two stations.
767 In Uccle, we have also tested the impact of smoothing the MAX-DOAS columns with the satellite averaging
768 kernels (similarly as for Xianghe), which turned out to be very small. The absence of any seasonal dependence in
769 the cities of Brussels (Uccle) and Bremen, in contrast to that observed (although limited) in Vienna and Athens,
770 is to some extent puzzling. One should keep in mind however that the glyoxal retrievals from MAX-DOAS
771 measurements are also challenging and it cannot be excluded that errors in ground-based data might also partly
772 contribute to the observed differences.

773 In Mohali and Pantnagar, glyoxal columns are much larger and the seasonal variability is driven by fire emissions
774 and meteorological factors such as the monsoon. At those two stations, the glyoxal seasonal variability is very
775 well reproduced by TROPOMI. In terms of absolute values, the TROPOMI columns agree reasonably well in
776 Mohali but, they significantly underestimate the (large) MAX-DOAS columns in Pantnagar. The reason why the
777 systematic satellite/ground-based bias is so different between those two stations is unclear. MAX-DOAS columns
778 are clearly higher in Pantnagar than in Mohali pointing either to possible local differences in air quality, not
779 reflected in the satellite data, or to inconsistencies in the ground-based data sets. Although the agreement is
780 excellent in Mohali, the typical behaviour is an underestimation of the columns by the satellites, as discussed
781 before. Note also that those sites are significantly contaminated by aerosols, which are neglected in the satellite
782 retrievals (apart from the stringent cloud filtering). MAX-DOAS data have also been analysed using very different
783 approaches, which may also cause differences. This calls for a more detailed analysis, which would require an
784 homogenization of the MAX-DOAS data treatment, a more sophisticated approach for the computation of the
785 satellite AMFs (e.g. with an explicit aerosol treatment) and possibly some independent information on the glyoxal
786 vertical distribution. This being said, the nice consistency in the glyoxal column seasonal variability by the
787 different systems is remarkable in itself. Table 4 provides an overview of the correlation coefficient between the
788 satellite and the MAX-DOAS glyoxal columns at all considered stations. For stations where the analysis was
789 possible for all satellite sensors, the correlation coefficient was found to be significantly better for TROPOMI
790 than for the other instruments. It is also clear that correlation coefficients are better for sites characterised by large
791 and highly variable glyoxal columns (e.g. Asian stations). Apart from the Bremen station where the negative bias
792 during winter leads to a low correlation coefficient, all other values are quite reasonable (between 0.61 and 0.87)
793 for TROPOMI. Table 4 also gives the mean bias as derived from the comparison of the satellite and MAX-DOAS
794 glyoxal column seasonal cycle as well as the standard deviation of the differences. As discussed above, the mean



795 differences are generally lower than 1×10^{14} molec/cm², except for high columns where differences are noticeably
 796 higher.



797

798 **Figure 16 :** Comparison of the monthly median glyoxal tropospheric vertical column seasonal cycle as retrieved from
 799 **TROPOMI** and **MXD** at four European stations (Bremen, Uccle, Vienna, Athens) and at two Indian stations (Mohali,
 800 **Pantnagar**). The error bars represent the interannual variability (25% and 75% percentiles based on the full time
 801 series available).

802

803 **Table 4 :** Correlation coefficients between the satellite and MAX-DOAS monthly median glyoxal tropospheric
 804 vertical columns as well as mean absolute difference and associated standard deviation at nine stations.

	Correlation coefficient Mean bias \pm standard deviation (10^{14} molec/cm ²)								
	Xianghe	Chiba	Phimai	Bremen	Uccle	Vienna	Athens	Mohali	Pantnagar
TROPOMI	0.87 -0.8 \pm 0.6	0.80 0.1 \pm 0.6	0.85 -2.0 \pm 0.8	0.13 -0.9 \pm 0.9	0.67 -0.5 \pm 0.7	0.73 -0.3 \pm 0.6	0.61 -0.4 \pm 0.6	0.70 0.6 \pm 0.9	0.78 -3.5 \pm 1.5
OMI (until 2013)	0.70 -0.7 \pm 1.3	0.32 -0.6 \pm 0.8	N/A						
GOME-2B (until 2016)	0.37 -0.9 \pm 0.9	0.66 0.0 \pm 0.7	0.88 -0.8 \pm 0.8						
GOME-2A	0.92 -0.8 \pm 0.4	0.58 -0.1 \pm 0.9	0.86 -1.1 \pm 0.8						



805 6. Conclusions

806 We presented the first global TROPOMI glyoxal tropospheric column product derived from three years (2018-
807 2020) of visible radiance measurements. The DOAS-based algorithm, which relies largely on previous
808 developments for heritage satellite nadir-viewing instruments, has been further improved in different aspects. In
809 particular, the use of additional pseudo cross-sections in the DOAS spectral fit allows mitigating the effect of the
810 instrumental spectral response function perturbations in case of scene brightness inhomogeneity, which otherwise
811 would lead to systematic biases in the retrieved glyoxal columns. This helps removing artefacts along the coasts
812 and reducing pseudo-noise in regions covered by persistent broken clouds. The glyoxal slant columns are also
813 empirically corrected for biases caused by the NO₂ misfit in case of strong absorption. Finally, the background
814 correction procedure has been optimized for the TROPOMI characteristics and the a priori glyoxal vertical
815 distribution, essential to the AMF computation, is now provided by the CTM MAGRITTE, an updated version of
816 the IMAGES model, running at the higher spatial resolution of 1°x1°. The glyoxal column retrievals have been
817 fully characterized with an error budget considering the different error components introduced in each of the
818 algorithm modules. This allows extending the glyoxal column data product with total random and systematic error
819 estimates provided for every observation, with corresponding averaging kernels and a priori profiles.

820 Glyoxal tropospheric columns have also been derived from data of the OMI, GOME-2A and GOME-2B satellite
821 instruments using retrieval baselines similar to the TROPOMI algorithm. An extensive inter-comparison of those
822 four data sets emphasised their excellent consistency with absolute mean glyoxal column differences found to be
823 generally lower than 0.5×10^{14} molec/cm². This demonstrates that glyoxal retrievals respond in the same manner
824 to our selection of settings for all nadir-viewing satellite instruments. Because of this sensitivity, the retrievals
825 may be easily impacted by spectral features caused by instrumental degradation. We have shown that the stability
826 of the OMI and GOME-2 data records is somewhat degraded after a few years of operations. Glyoxal retrievals
827 are characterized by a high level-of-noise, requiring significant spatio-temporal averaging to extract meaningful
828 signals. With both a much larger number of observations and a finer spatial resolution, TROPOMI outperforms
829 by far the previous instruments in its ability to provide high quality and detailed glyoxal fields.

830 Satellite observations have also been compared with a few independent MAX-DOAS data sets from stations
831 located in Asia and Europe. Owing to the scarcity of MAX-DOAS glyoxal data sets, especially covering several
832 seasons, this validation exercise is therefore unprecedented. Based on a thorough analysis at the Xianghe station
833 (China), where a 10-year time series of MAX-DOAS data is available, and on the comparison of seasonal cycles
834 at other stations, we conclude that satellite and MAX-DOAS instruments observe consistent glyoxal signals and
835 have similar intra-annual variations. This is reflected by the strong correlation coefficients, ranging between 0.61
836 and 0.87 for TROPOMI, with the exception of one mid-latitude station where the correlation is poorer. In general,
837 the satellite and MAX-DOAS columns also agree in absolute values with differences less than 1×10^{14} molec/cm²,
838 at least for stations with moderate columns. In Xianghe, we showed that the application of the satellite averaging
839 kernels to the MAX-DOAS data further reduces the mean differences. There are however two stations
840 (Phimai/Thailand and Pantnagar/India) where the satellite/MAX-DOAS bias is more significant, despite an
841 excellent agreement between the seasonal variations. The origin of this bias is not fully understood, but it is not
842 uncommon to have such biases in UV-Visible satellite retrievals for strongly polluted sites. In addition, we have
843 indications that the satellite observations are low-biased during wintertime at mid-high latitudes where both the



844 glyoxal signal is weak and the sensitivity to the boundary layer is reduced. The comparisons of OMI, GOME-2
845 and MAX-DOAS glyoxal columns also show reasonable agreement and similar intra-annual variability. Both the
846 correlation coefficients and the scatter of the satellite/ground differences were however less good than those of
847 TROPOMI. This points again to the better performance of TROPOMI for the detection of glyoxal from space and
848 to its enhanced capability at providing information on VOC emissions. For future work, it would be beneficial to
849 dedicate more efforts in the homogenization of the MAX-DOAS glyoxal retrievals in terms of both spectral
850 analysis and slant-to-vertical column conversion in order to strengthen their potential for the validation of satellite
851 data sets such as the one presented in this work.

852 **Data availability**

853 Access to TROPOMI glyoxal tropospheric column data is possible via the GLYRETRO website
854 (<https://glyretro.aeronomie.be/>), OMI glyoxal data can be obtained on request from the authors. Information to
855 download the GOME-2/Metop-A and GOME-2/Metop-B glyoxal data records is provided at
856 https://acsaf.org/datarecord_access.php.

857 **Author contributions**

858 CL is the main contributor to the study and led the writing of this paper. FH performed the validation exercise,
859 with support from MVR and LMAA. MVR, LMAA, AR, IDS, NT, JV, HY and JVG contributed to algorithm
860 and/or code development. TS and JFM provides the a priori modelled glyoxal profiles. PV and DL are responsible
861 for the production of the GOME-2 glyoxal operational data records. MVR, FH, LMAA, SFS, HI, VK, TW, VS,
862 TiW and PW contributed to operating the MAX-DOAS instruments, and to producing and providing glyoxal data.
863 CR supervised the study. All co-authors have been involved into the discussion of results and the writing of this
864 article.

865 **Competing interests**

866 The authors have the following competing interests: Thomas Wagner is chief-executive editor of AMT. Andreas
867 Richter is executive editor of AMT. Diego Loyola, Andreas Richter, Michel Van Roozendael and Thomas
868 Wagner act as associate editors for AMT.
869

870 **Acknowledgments**

871 This work contains modified Copernicus Sentinel-5 Precursor satellite data (2018-2020). It has been supported
872 by the European Space Agency via the preparation of the Level-2 Prototype Processor of the future Copernicus
873 Sentinel-5 satellite (contract #4000118463/16/NL/AI) and via the GLYRETRO project, part of the Sentinel-
874 5p+Innovation programme (contract #4000127610/19/I-NS). EUMETSAT, the Belgian Federal Science Policy
875 Office (BELSPO) and the German Aerospace Center (DLR) are acknowledged for their respective financial
876 support of the GOME-2 algorithmic developments through the AC SAF Continuous Development and Operations
877 Phase (CDOP-3) and the ProDEx B-ACSAF contribution to the ACSAF. The Vienna MAX-DOAS instrument is
878 part of the VINDOBONA project, which is funded by the Austrian Science Fund (FWF): I 2296-N29, the German
879 Science Foundation (DFG): Ri1800/6-1, and A1 Telekom Austria. IISER Mohali Atmospheric Chemistry Facility
880 is gratefully acknowledged for supporting the MAX-DOAS operations in Mohali, India. We thank Caroline Fayt
881 and Christian Herman from BIRA-IASB for maintaining the Uccle and Xianghe MAX-DOAS instruments.



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883 **References**

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