

1 **Observing downwind structures of urban HCHO plumes from space:**
2 **Implications to non-methane volatile organic compound emissions**

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16 **Key Points:**

- 17 • We show clear urban HCHO plumes from 16 cities over the globe by relating satellite
18 pixels with wind fields
- 19 • We obtain urban effective HCHO production rates by fitting the downwind structure of
20 HCHO plumes
- 21 • Satellite-based effective HCHO production rates provide potential measures of total non-
22 methane volatile organic compound emissions

23 **Abstract**

24 Non-methane volatile organic compounds (NMVOCs) have a significant impact on air quality in
25 urban areas. Detecting NMVOC emissions with its proxy HCHO on urban scales from space,
26 however, has been limited by the lack of discernible enhancement. Here we show clear urban
27 HCHO plumes from 16 cities over the globe by rotating TROPOspheric Monitoring Instrument
28 (TROPOMI) HCHO pixels according to wind directions. We fit the downwind structure of the
29 plumes with the exponentially modified Gaussian (EMG) approach to quantify urban HCHO
30 effective production rates between 7.0 mol s^{-1} and 88.5 mol s^{-1} . Our results are in line with total
31 NMVOC emissions from the EDGAR inventory ($r = 0.76$). Our work offers a new measure of
32 total NMVOC emissions from urban areas and highlights the potential of satellite HCHO data to
33 provide new information for monitoring urban air quality.

34 **Plain Language Summary**

35 Non-methane volatile organic compounds (NMVOCs) play an important role in urban air
36 quality. Formaldehyde (HCHO) satellite observations have been shown to be able to reliably
37 track and quantify NMVOC emissions at global and regional scales. Here, we use state-of-the-art
38 satellite sensors to quantify effective HCHO production rates in 16 global cities and further
39 constrain total NMVOC emissions. Our results are broadly consistent with current emissions
40 inventories, implying that satellites may be able to provide new information for urban air studies.

41 **1 Introduction**

42 Atmospheric formaldehyde (HCHO) is an intermediate produced via primary emission
43 and secondary formation from the oxidation of a range of volatile organic compounds (VOCs).
44 Therefore, the production rate of HCHO provides a potential constraint on the underlying VOC

45 emissions (Barkley et al., 2013; Bauwens et al., 2022; Shen et al., 2019; Zhu et al., 2014).
46 Previous field measurements show that anthropogenic non-methane VOC (NMVOC) emissions
47 are critical drivers of urban HCHO production rates (Liu et al., 2023; Zeng et al., 2019). Here,
48 we present the first attempt to apply satellite HCHO columns to estimate effective HCHO
49 production rates and to infer total anthropogenic NMVOC emissions in urban areas over the
50 globe by analyzing the downwind structures of their HCHO plumes.

51 Regional and local HCHO enhancements result from NMVOCs emitted by plants
52 (Barkley et al., 2013; Millet et al., 2006; Palmer et al., 2003; Wells et al., 2020; Wolfe et al.,
53 2016), fires (Alvarado et al., 2020; Cao et al., 2018; Holzinger et al., 1999; Yokelson et al.,
54 1999), and human activities (Bauwens et al., 2022; Pu et al., 2022; Shen et al., 2019; Sun et al.,
55 2021; Zhu et al., 2014; Zhu, Mickley, et al., 2017). In urban areas, the use of natural gas, diesel,
56 gasoline, and solid fuels results in direct emissions of HCHO and secondary production of
57 HCHO from various anthropogenic NMVOCs (Alzueta & Glarborg, 2003; Clairotte et al., 2013;
58 Green et al., 2021).

59 Satellites observe HCHO from space in a column manner. Previously, HCHO
60 tropospheric columns have been used in the inversion framework to constrain NMVOC
61 emissions from biogenic sources (Barkley et al., 2013; Millet et al., 2008; Millet et al., 2006;
62 Palmer et al., 2006; Wu et al., 2023) and fires (Cao et al., 2018; Fu et al., 2007; Gonzi et al.,
63 2011). However, applying HCHO columns to derive anthropogenic NMVOC emissions is
64 challenging due to the (1) high uncertainty in the *a priori* estimations (Huang et al., 2017; Zheng
65 et al., 2018), (2) lack of discernible enhancement on urban scales (Zhu et al., 2014), and (3)
66 highly nonlinear small-scale chemistry that makes using atmospheric chemistry transport models
67 challenging (Laughner & Cohen, 2019; Valin et al., 2013).

68 For gases (mainly NO₂ and SO₂) emitted from point sources (*e.g.*, megacities or power
69 plants), the combined analysis of satellite observations and wind fields reveals the downwind
70 decay of plumes and has been further used to estimate their lifetimes and emissions (Beirle et al.,
71 2011; de Foy et al., 2015; V. E. Fioletov et al., 2016; Goldberg et al., 2019; Lee et al., 2022; Lu
72 et al., 2015). However, similar observation-based approaches have long been recognized as
73 missing for NMVOCs, which are equally important for urban air pollution (von Schneidmesser
74 et al., 2023). In this study, we show evident downwind decay of urban plumes over the globe
75 with the state-of-the-art TROPOMI instrument (Veefkind et al., 2012) and the wind rotation
76 technique. By fitting HCHO plumes with the exponentially modified Gaussian (EMG) function,
77 we obtain the effective HCHO production rates and lifetimes, reflecting the emission and rapid
78 photochemical oxidation of NMVOCs.

79 **2 TROPOMI HCHO columns and wind rotation approach**

80 Onboard the Copernicus Sentinel-5 Precursor platform, TROPOMI is a nadir-viewing
81 spectrometer launched in October 2017, which scans the whole globe within a day at a local
82 passing time of 13:30 and a nadir resolution of 5.5 km × 3.5 km (7 km × 3.5 km before August
83 2019). It achieves a spectral resolution of 0.55 nm in the 328-359 nm band range where HCHO
84 retrieval is performed. We use 2019-2022 TROPOMI HCHO tropospheric vertical column
85 product (De Smedt et al., 2018), which has been thoroughly validated (Chan et al., 2020; De
86 Smedt et al., 2018; Vigouroux et al., 2020) and used to study NMVOC emissions (Pu et al.,
87 2022; Sun et al., 2021; Wang et al., 2022). To ensure data quality, we select level 2 pixels with
88 quality assurance (QA) value greater than 0.5, cloud fraction less than 0.3, and solar zenith angle
89 less than 60°.

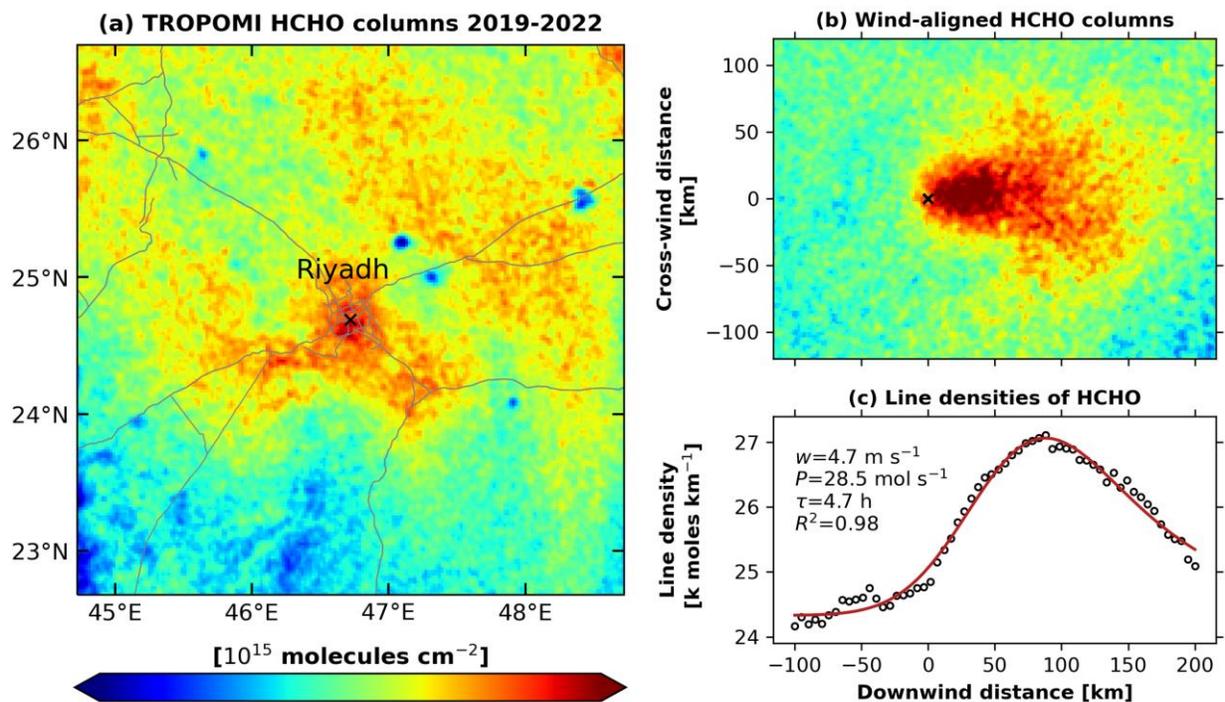
90 To investigate the downwind structures of urban HCHO plumes, we associate each pixel
91 with its wind direction and speed, sampled from the ECMWF Reanalysis v5 (ERA5) hourly data
92 (Hersbach et al., 2020). We use the average ERA5 wind fields in the bottom 5 levels (~ up to 1.0
93 km above sea level), following V. E. Fioletov et al. (2015). We then apply the wind rotation
94 technique (de Foy et al., 2015; V. E. Fioletov et al., 2015; Lu et al., 2015; Pommier et al., 2013;
95 Valin et al., 2013) to rotate each TROPOMI pixel around the city center (apparent source)
96 according to wind direction. Figure S1 illustrates the schematic of such a wind rotation approach.

97 First proposed by Valin et al. (2013) in their study of NO₂ urban plumes, the wind
98 rotation approach effectively redistributes satellite observations near the source along the
99 downwind direction. After rotation, all TROPOMI pixels have a common wind direction and can
100 be analyzed together, which helps us to accumulate a statistically significant TROPOMI HCHO
101 data set while preserving the upwind-downwind characteristics of each pixel. Another advantage
102 of wind rotation is that it makes the central source more pronounced while attenuating the signals
103 of the surrounding sources (V. E. Fioletov et al., 2015), which is particularly helpful for the
104 capture of HCHO urban plumes, as for HCHO the background levels are much higher and the
105 sources are less localized than NO₂ and SO₂.

106 **3 Observing and fitting urban HCHO plume: an example from Riyadh**

107 Our attempt starts with Riyadh (Saudi Arabia), one of the largest cities on the Arabian
108 Peninsula and is usually considered as an ideal place for satellite detection of urban plumes (*e.g.*,
109 NO₂ and CO) due to its isolated location, large emission, and frequent clear sky conditions
110 (Beirle et al., 2011; Lama et al., 2022; Valin et al., 2013). It is also an optimal spot to observe the
111 HCHO urban plume as it is surrounded by desert and has low biogenic VOC emissions. Figure
112 1a shows the 2019-2022 mean TROPOMI HCHO tropospheric columns around Riyadh with a

113 $0.02^\circ \times 0.02^\circ$ ($\sim 2 \text{ km} \times 2 \text{ km}$) resolution. The oversampling method we use is a weighted
 114 average of the satellite pixels on each grid, with weights obtained based on the overlap area of
 115 the pixels with the grid (Zhu, Jacob, et al., 2017). Wind rotation allows us to see a distinct urban
 116 HCHO plume above the regional background (Figure 1b). By integrating the two-dimensional
 117 HCHO plume (Figure 1b) along the cross-wind direction, we obtain the one-dimensional HCHO
 118 line densities, which exhibit a Gaussian shape and decay pattern (black circles in Figure 1c). We
 119 find that the maximum enhancement of HCHO ($\sim 27 \text{ k mol km}^{-1}$) occurs at about 75 km
 120 downwind of Riyadh, which is almost three times the distance of NO_2 maximum enhancement
 121 under fast wind conditions (Valin et al., 2013). This highlights the difference in lifetimes of
 122 HCHO and NO_2 , which implies the additional secondary production for HCHO from NMVOCs.



123

124 **Figure 1.** TROPOMI HCHO columns and downwind plume structure in Riyadh. (a) TROPOMI
 125 HCHO oversampled to $0.02^\circ \times 0.02^\circ$ ($\sim 2 \text{ km} \times 2 \text{ km}$) resolution from 2019 to 2022, with the

126 black cross marking the city center. Gray lines denote trunk roads and motorways. (b) Wind-
 127 aligned HCHO plume in Riyadh. (c) Line densities (black circles) of HCHO columns as a
 128 function of downwind distance from the city center. Each circle represents the TROPOMI
 129 HCHO line density integrated along the cross-wind direction (± 100 km). The red curve ($\Omega_{\text{line}}(x)$,
 130 see Section 3) is the exponentially modified Gaussian (EMG) fitting result, with the averaged
 131 wind speed (w) from ERA5 data, fitted effective HCHO production rate (P), and fitting
 132 determination coefficient (R^2) insert.

133 The exponential modified Gaussian (EMG) method has been widely applied in fitting the
 134 downwind plumes of NO_2 (Goldberg et al., 2019; Jin et al., 2021; Laughner & Cohen, 2019; Lu
 135 et al., 2015; Pommier, 2023) and SO_2 (Beirle et al., 2014; V. E. Fioletov et al., 2015; McLinden
 136 et al., 2016). This method assumes an approximate point source elevated from the background
 137 (V. Fioletov et al., 2022; Lange et al., 2022), which can be tested with the signal-to-noise (SNR)
 138 ratio that compares the upwind-downwind difference with satellite signals (McLinden et al.,
 139 2016; Pommier, 2023).

$$SNR = \frac{\Omega_d - \Omega_u}{\frac{\sigma_{\Omega_d}}{\sqrt{N_d}} + \frac{\sigma_{\Omega_u}}{\sqrt{N_u}}} \quad (1)$$

140 where Ω_d and Ω_u is the average HCHO column in downwind and upwind regions at the same
 141 distance from the center (Figure S2); σ_{Ω_d} , σ_{Ω_u} , N_d , and N_u is the standard deviation and number
 142 of observations in the two regions, respectively. To ensure sufficient contrast to the background,
 143 we set an SNR threshold of 10.0 to determine an approximate point source for HCHO,
 144 considering the lifetime of HCHO (few hours, similar to NO_2 and SO_2), resolution of
 145 TROPOMI, and size of the sources observed in each city. For Riyadh, the SNR value is 15.6.

146 The EMG method to fit HCHO line densities $\Omega_{\text{line}}(x)$ (Figure 1c) is:

$$\Omega_{\text{line}}(x|\mu, \sigma, x_0, \alpha, B) = \alpha \cdot \left[\frac{1}{x_0} \exp\left(\frac{\mu}{x_0} + \frac{\sigma^2}{2x_0^2} - \frac{x}{x_0}\right) \Phi\left(\frac{x - \mu}{\sigma} + \frac{\sigma}{x_0}\right) \right] + B \quad (2)$$

147 where α (mol) is a scale factor of the total number of HCHO molecules observed near the
 148 hotspot, elevated from the background (B , mol km⁻¹); μ (km) is the location of the point source
 149 relative to the urban center (defined as $x = 0$); x_0 (km) is the e -folding distance downwind; σ
 150 (km) is the standard deviation of the Gaussian function; and Φ is the cumulative distribution of
 151 exponential function.

152 Similar to studies on NO₂ and SO₂ point source emissions, we define an effective lifetime
 153 of HCHO (τ^*) as:

$$\tau^* = x_0 / w \quad (3)$$

154 Here τ^* (hour) represents the effective mean lifetime of HCHO within the fitting domain from an
 155 approximate point source, encapsulating the effects of primary emission, secondary production,
 156 loss, and transport. w (4.7 m s⁻¹) is the effective wind speed of the study domain according to
 157 ERA5 wind fields. Further, the effective HCHO production rate P (mol s⁻¹) is defined as:

$$P = \alpha / \tau^* \quad (4)$$

158 which includes both primary HCHO emitted in the city and secondary HCHO produced within
 159 the downwind plume.

160 For Riyadh, the fitted line densities are close to TROPOMI observations with a
 161 determination coefficient (R^2) of 0.98 (Figure 1c), an effective lifetime of HCHO (τ^*) of 4.3 \pm
 162 1.1 hours (95% confidence interval), and an effective HCHO production rate (P) of 33.1 \pm 3.6
 163 mol s⁻¹. The fitted background (B) is 24.4 \pm 0.1 k mol km⁻¹, corresponding to a column density of

164 7.3×10^{15} molecules cm^{-2} in the fitting domain, which we attribute to the oxidation of regional
165 biogenic (*e.g.*, isoprene) and long-lived VOCs (*e.g.*, methane). Here, we refer to Beirle et al.
166 (2011) and Lu et al. (2015) to quantify the uncertainties of our results (Text S2).

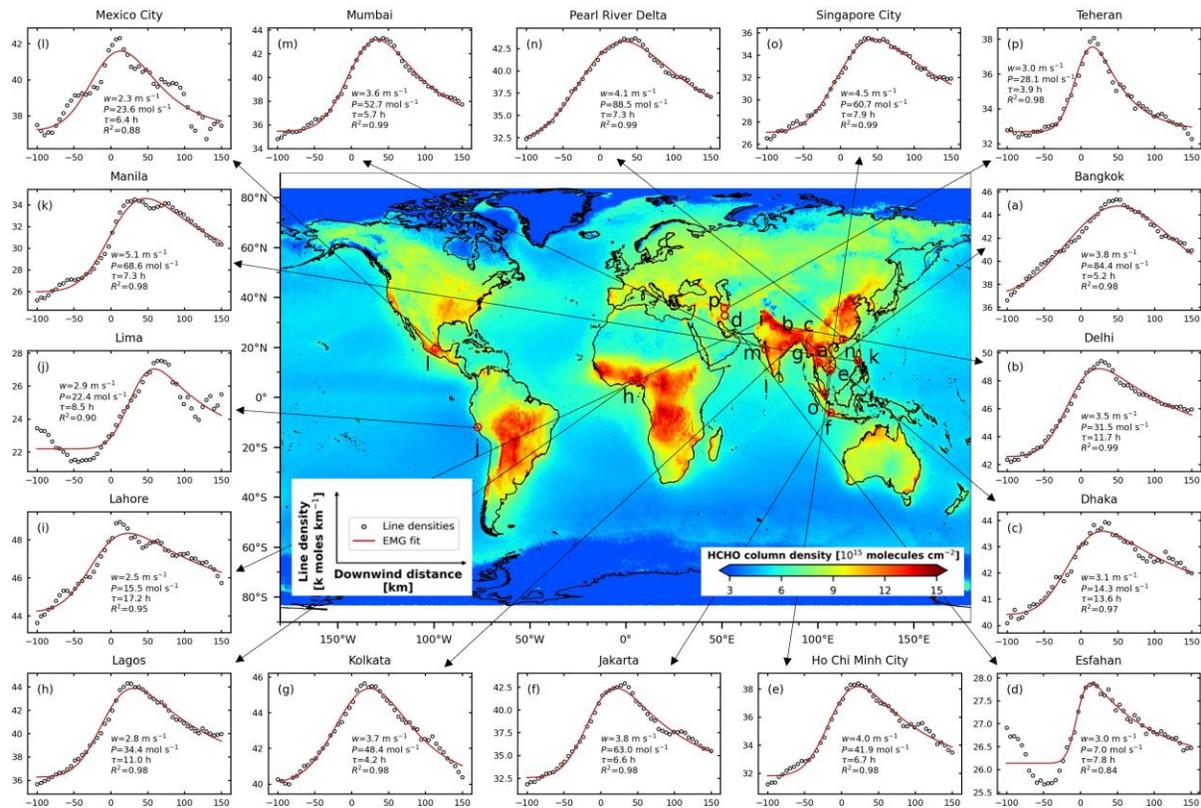
167 **4 Downwind structures of urban HCHO plumes over the globe**

168 As demonstrated in Riyadh (Figure 1b and 1c), the wind rotation approach enables
169 detection of urban HCHO plumes. Based on this, we extend our analysis globally by focusing on
170 55 cities or urban agglomerations with populations over 5 million and another 11 cities with
171 visible HCHO enhancements. Table S1 lists those 66 cities or urban agglomerations, among
172 which 25 satisfy the point source criterion (*i.e.*, $SNR > 10.0$).

173 We then apply the EMG method for each approximate point source candidate city in a
174 200 km by 250 km (± 100 km cross-wind, 100 km upwind, and 150 km downwind) domain. This
175 domain size is selected to minimize interference from surrounding sources (biogenic and
176 anthropogenic) while retaining enough satellite pixels. Following Jin et al. (2021) and Laughner
177 & Cohen (2019), we set additional criteria to obtain reasonable EMG fitting: (1) $R^2 > 0.8$, which
178 ensures the fitted EMG curve is close to the observations; (2) $x_0 > \sigma$, which requires emission
179 width shorter than the e -folding distance to avoid the case that emission shape confounds with
180 HCHO decay structure; and (3) $(150 \text{ km} - \mu) / w > \tau^*$, which states the plume residence time
181 should be longer than the effective HCHO lifetime to reduce EMG fitting uncertainty. Table S1
182 provides whether each criterion is valid for the approximate point source candidate cities.

183 Figure 2 shows downwind structures of the resulting plumes in 16 cities or urban
184 agglomerations, with wind-aligned HCHO plumes provided in Figure S3. Table S2 summarizes
185 the corresponding EMG fitting results. The fitted effective production rate of HCHO (P) ranges

186 from 7.0 mol km^{-1} (Esfahan) to 88.5 mol km^{-1} (Pearl River Delta), with background (B) ranges
 187 from $22.2 \text{ k mol km}^{-1}$ to $44.2 \text{ k mol km}^{-1}$ (6.6×10^{15} molecules cm^{-2} to 13.2×10^{15} molecules cm^{-2}
 188 ², Table S2). The effective lifetime (τ^*) is between 4.0 hours (Teheran) and 17.2 (Lahore) hours.

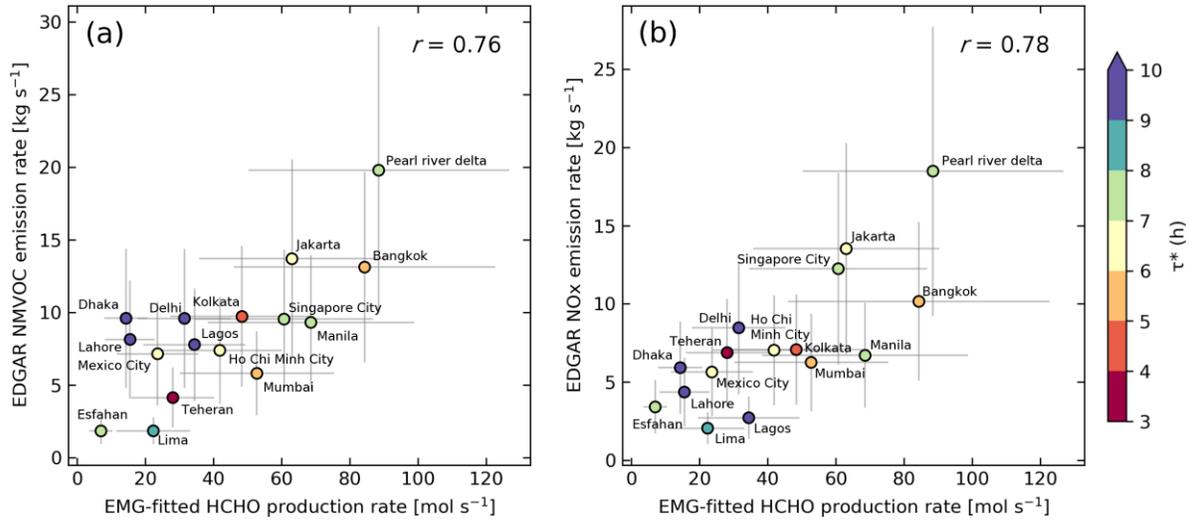


189
 190 **Figure 2.** Global urban HCHO hotspots and their downwind plume structures. The center panel
 191 shows the mean TROPOMI tropospheric HCHO columns from 2019 to 2022 at a resolution of
 192 $0.1^\circ \times 0.1^\circ$ ($\sim 10 \text{ km} \times 10 \text{ km}$), with HCHO hotspots circled in red. The black arrow points to the
 193 observed (black circles) and fitted (red curve) line densities for each approximate point source
 194 city (panel a-p). The horizontal coordinate is the downwind distance (km) from the

195 corresponding city center. The vertical coordinate is HCHO line density (k mol km^{-1}), integrated
196 along the cross-wind direction.

197 As shown in Figure 3, our effective production rates of HCHO (P) are generally in line
198 with local total anthropogenic NMVOC (panel a; $r = 0.76$) emissions from EDGAR (v6.1)
199 within a 100 km radius of the city center. Within the fitting domains of the 16 cities or urban
200 agglomerations in Figure 3, biogenic isoprene emission (MEGAN v2.1 run for 2019; Guenther et
201 al., 2012) accounts for on average 25% of the total NMVOC emissions, arguing for an
202 anthropogenic dominated origin of HCHO. This could be further backed up by the consistency (r
203 $= 0.78$) between effective HCHO production rates and anthropogenic nitrogen oxides (NO_x ;
204 Figure 3 panel b) emissions.

205 We acknowledge the spatial heterogeneity of biogenic NMVOC emissions over hundreds
206 of kilometers around the city, which may introduce uncertainties to the fitting results. In
207 addition, the temporal difference between TROPOMI overpass time (13:30 local time) and
208 EDGAR inventory (24-hour average for the year 2018) may be another source of uncertainties.
209 Nevertheless, the broad agreement with EDGAR inventory corroborates the reliability of our
210 approach, suggesting the effective production rate of HCHO could be a potential measure of total
211 anthropogenic NMVOC emissions in urban areas despite NMVOCs contributing to HCHO
212 differently under various OH levels.



213

214 **Figure 3.** Comparison between EMG-fitted effective HCHO production rates with total
 215 anthropogenic NMVOC (panel a) nitrogen oxides (NO_x; panel b) from EDGAR. The fitted rates
 216 are from the EMG fitting results (Section 4). Each point represents a city in Figure 2, colored by
 217 its fitted effective lifetime (τ^*) that saturates at 10 hours. Error bars show the total uncertainties
 218 of P_{HCHO} (Text S1), whereas EDGAR uncertainties are set to be 50% (Beirle et al., 2011). The
 219 Pearson correlation coefficient (r) is also inserted.

220 Previous studies report that the EMG method accurately estimates emissions, but the
 221 effective lifetime is not a reliable measure of a gas chemical lifetime due to plume meandering
 222 and grid resolution (de Foy et al., 2015). In addition, secondary production also complicates the
 223 effective lifetime of HCHO, along with physical diffusion and chemical losses in urban plumes.
 224 That being said, τ^* depends mainly on the photolysis rate and OH concentration. If photolysis
 225 rates are known, one could determine the OH level of the urban plume (Liao et al., 2021) in a
 226 similar rationale to the study of NO₂ lifetimes (de Foy et al., 2015; Laughner & Cohen, 2019;
 227 Valin et al., 2013). Such information may help us better quantify urban atmospheric oxidation
 228 levels through satellite remote sensing.

229 **5 Conclusion**

230 We have used TROPOMI satellite observations and ERA5 wind fields to detect urban
231 HCHO plumes from 16 cities over the globe. By fitting the downwind structure of the plumes,
232 we quantify effective HCHO production rates in urban areas, which are in line with total non-
233 methane volatile organic compound (NMVOC) emissions from the bottom-up inventory. Our
234 work shows the potential of satellite HCHO columns in providing new information for urban air
235 quality studies.

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250 German Aerospace Centre (DLR).

251 **Open Research**

252 We gratefully acknowledge the dataset of TROPOMI HCHO product
253 (<https://doi.org/10.5270/S5P-tjlxfd2>), the ERA5 dataset
254 (<https://cds.climate.copernicus.eu/#!/search?text=ERA5&type=dataset>), the EDGAR v4.3.2
255 (https://edgar.jrc.ec.europa.eu/dataset_ap432_VOC_spec), EDGAR v6.1
256 (https://edgar.jrc.ec.europa.eu/dataset_ap61), and population (www.geonames.org) database. The
257 oversampling code is available at: <https://github.com/zhu-group/RegridPixels>.

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