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Sara Martínez-Alonso, Pepijn Veefkind, Barbara Klara Dix, Benjamin Gaubert, Nicolas Theys, et al.. TROPOMI-derived NO2 emissions from copper/cobalt mining and other industrial activities in the Copperbelt (DRC and Zambia). 2023. hal-04235277

HAL Id: hal-04235277 https://cnrs.hal.science/hal-04235277

Preprint submitted on 10 Oct 2023

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TROPOMI-derived NO₂ emissions from copper/cobalt mining and other industrial activities in the Copperbelt (DRC and Zambia)

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16 Key Points:

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17	· We quantified annual 2019-2022 TROPOMI-derived NO ₂ emissions from six point
18	sources in the Copperbelt, despite high background emissions.
19	\cdot Annual TROPOMI-derived NO ₂ emissions from these point sources are strongly
20	correlated with annual mine/oil refinery production.
21	\cdot Lack of elevated SO ₂ at these point sources is consistent with SO ₂ capture and

²² production of sulfuric acid, a profitable byproduct.

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23 Abstract

- ²⁴ We have analyzed TROPOMI data over the Copperbelt mining region (Democratic Re-
- ²⁵ public of Congo and Zambia). Despite high background values, we find that annual 2019-
- 26 2022 means of TROPOMI NO₂ show local enhancements consistent with six point sources
- 27 (mines and cities) where high-emission industrial activities take place. We have quan-
- tified annual NO₂ emissions for the six sources, identified temporal trends in these emis-
- ²⁹ sions, and found strong correlations with mine/refinery production data. CAMS-GLOB-
- ANT v5 inventory emissions are lower than TROPOMI-derived emissions by 61-96 %
- and lack the temporal trends observed in TROPOMI and mine/oil refinery production.
- ³² Lack of TROPOMI SO₂ enhancements over the point sources analyzed indicates SO₂
- ³³ capture and transformation into sulfuric acid, a profitable byproduct. These results demon-
- 34 strate the potential for satellite monitoring of mining/oil refining activity which impacts
- the air quality of local communities. This is particularly important for Africa, where mining is increasing aggressively.
- ³⁶ Ing is increasing aggressivery.

Plain Language Summary

We show that air pollution from copper/cobalt mines and oil refineries can be iden-38 tified and measured from satellite, even in the presence of high background pollution from 39 biomass burning and other sources; our findings may apply as well to other industries 40 that consume large quantities of fossil fuels. This is important for monitoring the air qual-41 ity of local communities, particularly when these industrial activities take place in close 42 proximity to population centers, as is the case in the Copperbelt and, in general, in other 43 African regions where mining and related industrial activities proliferate without suffi-44 cient air quality monitoring. Additionally, we show for the first time that the amount 45 of air pollution measured by TROPOMI is strongly correlated with mine/refinery pro-46 duction. Studies like this can be used to estimate mine/refinery production before com-47 panies release their annual reports or (as is the case with non-publicly traded compa-48 nies) in the absence of such reports. Insufficient emissions from mines claiming high pro-49 duction may be indicative of production from a different source. Thus, this method may 50 help improve the traceability of minerals extracted in conflict areas and smuggled into 51 the global supply chain despite existing traceability and tagging schemes. 52

53 **1 Introduction**

The Copperbelt, a mining region straddling the DRC (Democratic Republic of Congo) 54 and Zambia, is currently of great strategic interest because it is the world's largest cobalt 55 producer and holds almost half of the world reserves (Shedd, 2022). Cobalt production 56 in the Copperbelt (mostly in the DRC) has increased ~600% between 1990 and 2021 57 (U.S. Bureau of Mines, 1993; Shedd, 2022), driven by its use in lithium-ion batteries which 58 power mobile phones, laptops, and electric cars. Access to the Copperbelt's cobalt is be-59 coming a matter of national and global energy security (Gulley, 2022). Cobalt is, how-60 ever, a byproduct of copper mining; copper is the main ore (by volume) extracted in the 61 Copperbelt. Previous studies have documented the impact of cobalt and/or copper min-62 ing in the region's soils and water (Atibu et al., 2016), land use (Mwitwa et al., 2012), 63 and neonatal health (Kayembe-Kitenge et al., 2019; Van Brusselen et al., 2020). The im-64 pact on local air quality remained unknown. Here we quantify the effect of increasing 65 mining activity on the air quality of this region using TROPOMI (TROPOspheric Mon-66 67 itoring Instrument) satellite measurements (Veefkind et al., 2012) of NO₂ (nitrogen dioxide); TROPOMI SO₂ (sulfur dioxide) is also analyzed. Both gases are atmospheric pol-68 lutants (World Health Organization, 2021) relevant to air quality monitoring and fore-69 casting. They are also considered short-lived climate forcers, important for understand-70 ing climate (Myhre et al., 2013). 71

 NO_x (NO₂ + NO, two species closely intertwined by oxidation and reduction re-72 actions), has both anthropogenic (fossil fuel combustion, biomass burning) and natural 73 (microbial activity in soils, lightning, wildfires) sources. Mining-related NO_x is produced 74 by high-temperature combustion of fuel used by trucks and other heavy machinery as 75 well as by electric generators. The main sink of NO_x is the hydroxyl radical (OH), with 76 which it reacts within hours in the presence of light. NO_x has a negative impact on air 77 quality, both directly and as a precursor to tropospheric ozone and particulate matter. 78 It is damaging to human health (affecting mostly the respiratory system) and crops, and 79 contributes to the formation of smog and acid rain. Hereafter we discuss NO_{2} , the NO_{x} 80 component measurable from satellite. 81

Measuring global and regional NO2 was made possible by satellite instruments such 82 as GOME (Global Ozone Monitoring Experiment), SCIAMACHY (SCanning Imaging 83 Absorption SpectroMeter for Atmospheric CHartographY), OMI (Ozone Monitoring In-84 strument), and GOME-2, (Leue et al., 2001; Richter et al., 2005; Beirle et al., 2011; Richter 85 et al., 2011). Labzovskii et al. (2022) reported regional-scale correlation between OMI 86 NO₂ column values from heavy industry, including mining, and a coal production inter-87 annual variability index for the Siberian Kuzbass Basin. Thanks to its higher spatial resolution, TROPOMI allows for the measurement of NO_2 over smaller domains such as 89 gas and oil fields (Dix et al., 2022), cities (Goldberg et al., 2019; Pommier, 2022; de Foy 90 & Schauer, 2022), and power plants (Beirle et al., 2019, 2021; Goldberg et al., 2019; Dix 91 et al., 2022; de Foy & Schauer, 2022). 92

 SO_2 results from both anthropogenic (e.g., coal combustion, smelting of sulfur-rich 93 ores) and natural (volcanism, marine biological processes) sources. It contributes to acid 94 rain and particle formation. Exposure to SO2 is harmful to human health, damages fo-95 liage, and impedes plant growth. Previous studies showed that SO₂ emissions could be 96 estimated using satellite data from TOMS (Total Ozone Mapping Spectrometer), GOME, 97 OMI, SCIAMACHY, and OMPS (Ozone Mapping and Profiler Suite) (Krueger, 1983; 98 Carn et al., 2007; V. E. Fioletov et al., 2013; Zhang et al., 2017). V. Fioletov et al. (2020, QC 2023) reported TROPOMI-based emission estimates of SO_2 from power plants, volca-100 noes, oil and gas fields, and smelters. 101

We show that copper/cobalt mining-related activities, among others, can be identified and their NO₂ emissions quantified based on TROPOMI data even in the presence of high background values from biomass burning and other sources (BIRA-IASB, 2021). Additionally, we identify inter-annual trends in TROPOMI-derived NO₂ emissions that are strongly correlated with mining and oil refinery production. Next we describe the datasets (Sect. 2) and methodology (Sect. 3) used in this study, we present our results (Sect. 4), and discuss their relevance (Sect. 5). Conclusions are offered in Sect. 6.

109 2 Datasets

TROPOMI, onboard the European Space Agency's Sentinel-5 Precursor satellite 110 (Veefkind et al., 2012), provides quasi-global daily coverage at high spatial resolution (3.5 111 $x 5.5 \text{ km}^2$ for our species of interest). This is a nadir-viewing imaging spectrometer in 112 a sun-synchronous orbit at 824 km of altitude, with 13:30 LST Equator-crossing time, 113 and 2600 km swath width. TROPOMI measures radiances in the ultraviolet, visible, and 114 reflected infrared, from which concentrations of trace gases as well as cloud and aerosol 115 properties are derived. Here we focus on TROPOMI measurements of tropospheric NO₂ 116 and SO_2 , two pollutants produced by mining-related activities. NO₂ is retrieved from 117 TROPOMI radiance measurements in the visible portion of the spectrum (400-496 nm) 118 (van Geffen et al., 2020; H. J. Eskes & Eichmann, 2021; H. Eskes et al., 2022). We used 119 daily TROPOMI NO₂ tropospheric column data from version 2 for the period between 120 1 January 2019 and 31 December 2022. We also analyzed TROPOMI SO₂ data retrieved 121 with the Covariance-Based Retrieval Algorithm (COBRA, Theys et al. (2021)) from ultraviolet-122

visible radiances (310.5-326 nm) (Theys, 2022) for the 1 January 2019 - 31 July 2022 period; more recent data were unavailable at the time of writing.

Meteorological information needed to derive emissions from TROPOMI NO₂ VCD (vertical column density) was obtained from reanalysis data. By combining measurements and model results, reanalyses datasets provide consistent and gapless global coverage of essential climate variables. We used hourly ERA5 (ECMWF Reanalysis v5) data (Hersbach et al., 2020) provided at 0.25° x 0.25° resolution and generated by the Copernicus Climate Change Service at the European Center for Medium-Range Weather Forecasts.

¹³¹ Due to the unavailability of ground measurements, we compared inventory data to TROPOMIderived NO2 emissions. Emission inventories are compilations of amounts of air pollutants released into the atmosphere, segregated by source and time period. We used 2019-2021 data from the CAMS-GLOB-ANT (Copernicus Atmosphere Monitoring Service Global Anthropogenic) emissions inventory version 5, an extension of version 4.2 (Granier et al., 2019). Inventory emissions for 2022 were unavailable at the time of writing. We focused on monthly NOx emissions, provided at $0.1^{\circ} \times 0.1^{\circ}$ resolution.

Mine production data were obtained from the annual reports of publicly traded min ing companies; these reports are mandated by official regulatory bodies such as the Se curities and Exchange Commission in the United States and the Securities and Futures
 Commission in Hong Kong. Private mining companies are not required to disclose their
 production data. The specifics of the information available in these reports varies greatly.

143 **3 Methodology**

To identify potential emission point sources such as mines, we produced annual means from daily TROPOMI NO₂ VCD for the Copperbelt study area (-10.5°N to -13.5°N, 24.5°E to 29.5°E). Temporal averaging enhances the signal from constantly emitting sources while dampening more sporadic, background emissions from biomass burning, soils, and lightning.

Once the point sources were identified, we calculated daily TROPOMI-derived NO₂ 1/0 emissions for the study area using the divergence method (Beirle et al., 2019; Dix et al., 150 2022) with some modifications described below. This method derives emission based on 151 a divergence term and a sink term, which account for wind dispersion effects and NO_2 152 depletion by OH, respectively. A detailed description of the derivation can be found in 153 Dix et al. (2022). To calculate daily emissions we regridded daily VCD values to a com-154 mon 0.025° x 0.025° grid. We filtered TROPOMI measurements to avoid clouds, errors, 155 and problematic retrievals by using only those with quality assurance value ≥ 0.75 (H. Es-156 kes et al., 2022); retrievals with solar zenith angle $>60^{\circ}$ were rejected. Hourly values 157 of fields (longitudinal and latitudinal horizontal wind at 100 m from the ground, pres-158 sure, and temperature) required for the emissions calculation were obtained from ERA5 159 reanalysis. All ERA5 fields were resampled to 0.025° x 0.025° spatial resolution and in-160 terpolated to the passing time of the closest (spatially and temporally) TROPOMI ob-161 servation. Fields provided on pressure levels were interpolated to an altitude of 100 m 162 above the ground. In our implementation, OH lifetime was calculated for each data point 163 based on the solar zenith angle of the closest TROPOMI retrieval. Daily emissions were averaged into annual means. 165

To calculate the actual NO₂ emissions released from each point source, background NO₂ emissions must be quantified and removed from the raw (non-background corrected) emissions. Several background removal approaches are possible: Beirle et al. (2019) subtracted from the emissions their 10th percentile value, Dix et al. (2022) removed the mode of a Gaussian curve fit to them. We find that the former is better suited to study regions with homogeneous background emissions and that results from the latter are highly dependent on how the Gaussian curve is defined. Similarly, statistics derived from two-dimensional

Gaussians fitted to the point source emissions (Beirle et al., 2019) were, in our case, highly 173 dependent on location and size of the area selected to perform the fit. Our approach con-174 sisted of calculating annual statistics (mean and standard deviation) for each point source 175 plume as well as for its local background area: $a \sim 1^{\circ} \times 1^{\circ}$ region surrounding each point 176 source, excluding the point source plume. The location and extent of the point source 177 plumes were identified based on an empirical threshold $(0.37 \text{ kg km}^{-2} \text{ h}^{-1})$ applied to 178 mean raw emissions from the 2019-2022 period. Annual means of background-removed 179 emissions were calculated for each point source by subtracting its mean background value 180 from its mean raw emission value. 181

Dix et al. (2022) described in detail the effects of individual parameters (e.g., background correction, wind level, wind data source, OH lifetime, VCD thickness) on the results obtained using this method. We investigated ERA5 wind data uncertainty effects by using the spread of its wind field 10 member ensemble to perturb 2020 NO₂ emissions. The spread provides an estimate of relative, random uncertainty; ensemble, mean, and spread are part of the ERA5 dataset. The results show that ERA5 wind uncertainty produces small changes in raw NO₂ emission (< 4 %; Table S1 and Fig. S1).

189 **4 Results**

Six distinct point sources are clearly visible in the mean 2019-2022 TROPOMI NO_2 190 VCD map (Fig. 1; Fig. S2 shows annual mean VCD maps). Four of the point sources 191 correspond to large copper (Sentinel, a.k.a. Trident; Lubumbashi; Kansanshi) or copper-192 cobalt (Kolwezi and adjacent Katanga) open-pit mines. The remaining two point sources 193 coincide with cities (Lubumbashi and Ndola) where we infer that high-emission indus-194 trial activities take place, as explained below. Latitudinal profiles of mean annual VCD 195 across each point source (Fig. 1) show that background NO_2 remains nearly constant year to year while the point sources (well defined, narrow peaks of fixed location) vary in magnitude. TROPOMI emission results (Table 1 and Fig. S3) reinforce these obser-198 vations. TROPOMI NO₂ mostly increased with time at all the point sources but one, 199 Ndola, where background-removed emissions decreased by > 70 % between 2019 and 2022. 200

To understand these inter-annual trends in NO₂ emissions, we compared background-20 removed TROPOMI-derived emissions to mine production data where available (Table 202 1 and Fig. 2). Most of the energy consumed in copper or copper-cobalt mining, includ-203 ing electricity, is generated by diesel fuel combustion. Mining equipment consumes ~ 60 204 % of the total energy; comminution \sim 36 %; flotation, filtering, and drying \sim 4 % (Allen, 205 2021). Limited data relevant to energy consumption is provided in mining company re-206 ports; we found the best proxy for energy (i.e., diesel) consumed to be amount of ore and 207 waste mined. Panels b, c, and d in Fig. 2 show strong positive correlation between an-208 nual values of total ore plus waste mined versus NO₂ emissions for the Sentinel, Lumwana, 209 and Kansanshi mines (R = 0.84, 0.74, and 0.79, respectively). Amount of copper pro-210 duced (highly dependent on ore grade, among other factors, and thus, a less-desirable 211 proxy) was used for the Kolwezi-Katanga mines (Fig. 2.a) for lack of ore and waste data 212 (R = 0.84).213

The remaining two point sources coincide with some of the largest cities in the study 214 area: Lubumbashi (DRC, population > 2.6 x 10⁶) and Ndola (Zambia, population > 0.5 215 \times 10⁶). To discard the hypothesis that their emissions are due to urban activity alone, 216 we quantified NO₂ emissions for two additional cities of similar size: Mbuji-Mayi (DRC, 217 population > 2.7 x 10⁶), located 750 km northwest of Lubumbashi; and Kitwe (Zambia, 218 population > 0.7 x 10⁶), 50 km northwest of Ndola (labeled g in Fig. 1 map). (Popu-219 lation data are for 2022 urban areas (Population Stat, 2023).) The results (Fig. 1, Ta-220 ble S2, and Fig. S4) show that neither the magnitude of VCD or emissions, the spatial 221 extent of the plumes, or the temporal emission trends from Lubumbashi and Ndola can 222 be explained by urban activity alone. Background-removed NO₂ emissions from Lubum-223

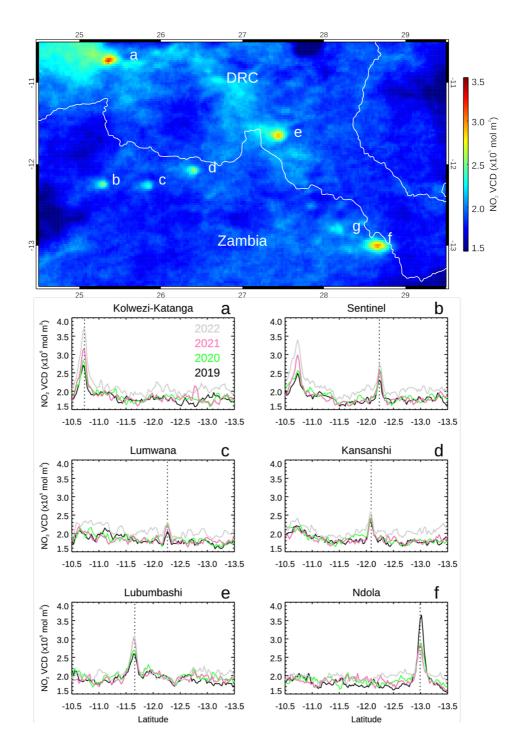


Figure 1. (Top) 2019-2022 mean of TROPOMI NO₂ VCD for the Copperbelt study region. Labels a through f show the six point sources analyzed. Label g shows the location of Kitwe City. White lines indicate country borders. (Bottom) Annual means of TROPOMI NO₂ VCD along latitudinal profiles centered at each of the six point sources, shown by vertical dotted lines. Labels as above. Profiles are color-coded according to year.

Table 1. Production and NO_2 emissions (TROPOMI background-removed, TROPOMI background, inventory) for six Copperbelt point sources. Ore, waste, and copper produced are in ktonnes; copper grade in percentage. Kansanshi reports provide separate copper grade values for sulfide, mixed, and oxide ores. Throughput (slag processed in Lubumbashi, crude refined in Ndola) are in ktonnes. All emissions and their standard deviation values (in parenthesis) are in kg km² h⁻¹.

		2019	2020	2021	2022
	Ore, Waste	-, -	-, -	-, -	-, -
	Grade	4.2, -	4.17, -	2.9, -	-, -
	Copper	84.3, 234.5	114.3, 270.7	121.1, 264.4	-, 220.1
Kolwezi ^a , Katanga ^b	TROPOMI	0.19 (0.10)	0.24 (0.12)	0.31 (0.15)	0.41 (0.17)
0	Background	0.23 (0.08)	0.23 (0.10)	0.22 (0.10)	0.26 (0.10)
	Inventory	0.03	0.03	0.03	-
	Ore, Waste	50263, 92826	60098, 97970	57380, 102445	56219, 95335
	Grade	0.5	0.49	0.47	0.46
	Copper	220.006	251.216	232.688	242.451
Sentinel ^c	TROPOMI	0.23 (0.11)	0.27 (0.12)	0.33 (0.10)	0.26 (0.11)
	Background	0.20 (0.07)	0.22 (0.08)	0.21 (0.07)	0.24(0.07)
	Inventory	0.01	0.01	0.01	-
	Ore, Waste	23230, 62837	26880, 73480	33510, 65499	20277, 78063
	Grade	0.47	0.52	0.46	0.52
	Copper	107.955	125.191	109.769	121.109
Lumwana ^d	TROPOMI	0.20 (0.06)	0.29 (0.08)	0.27 (0.09)	0.21 (0.06)
	Background	0.21 (0.07)	0.22 (0.07)	0.21 (0.07)	0.24 (0.07)
	Inventory	0.04	0.04	0.04	-
	Ore, Waste	36325, 52768	34423, 61972	35142, 69758	28205, 75878
	Grade	0.89, 1.05, 1.12	0.83, 1.00, 0.93	0.88, 0.96, 0.72	0.71, 0.63, 0.57
	Copper	232.243	221.487	202.159	146.282
Kansanshi ^c	TROPOMI	0.22 (0.07)	0.31 (0.09)	0.33 (0.11)	0.28 (0.10)
	Background	0.22 (0.07)	0.23 (0.08)	0.22 (0.07)	0.24(0.07)
	Inventory	0.08	0.08	0.08	-
	Throughput	-	-	255.229	-
Lubumbashi ^e	TROPONI	0.23 (0.10)	0.23 (0.10)	0.30 (0.16)	0.28 (0.17)
	Background	0.25 (0.08)	0.26 (0.08)	0.25 (0.08)	0.26 (0.08)
	Inventory	0.25	0.26	0.26	-
	Throughput	700.277	372.384	56.672	0^g
NJalaf	TROPOMI	0.58 (0.30)	0.32 (0.10)	0.26 (0.13)	0.17 (0.10)
Ndola ^f	Background	0.20 (0.09)	0.21 (0.09)	0.21 (0.08)	0.23 (0.08)
	Inventory	0.22	0.22	0.22	-

-7-

^aZijin (2023). ^bGlencore (2023). ^cFirst Quantum (2023a). ^dBarrick (2023). ^eSTL (2023). ^fMwila et al. (2022). ^gInactive.

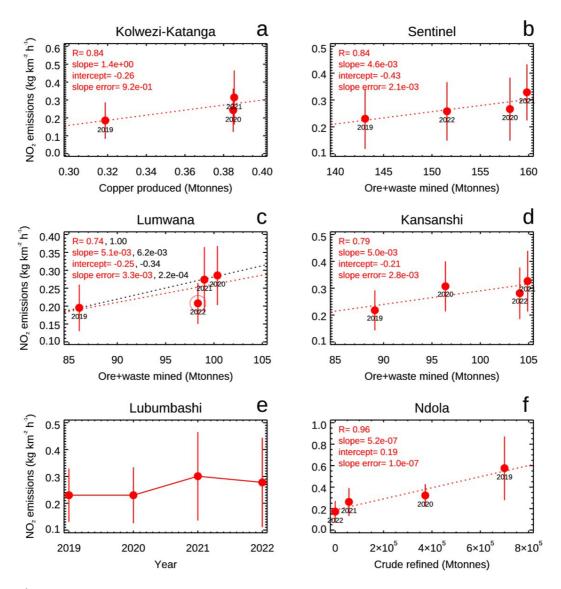


Figure 2. TROPOMI-derived, background-removed NO_2 emissions and their standard deviation versus production, except (e), where production data is unavailable. (c) Statistical values and fit line in black exclude 2022 values; see text for details.

bashi are higher than those from Mbuji-Mayi by 90 %, even though the population of 224 the former is lower by 2 %. Similarly, while its population is lower by 27 %, emissions 225 from Ndola surpass those from Kitwe by 40-80 %, depending on the year, and display 226 inter-annual variations which do not correlate with changes in population. These find-227 ings are consistent with our hypothesis that emissions from Lubumbashi and Ndola are not the result of urban activity alone. We researched other possible origins for the ex-22 cess emissions observed at these two point sources. We hypothesize that Lubumbashi's 230 high NO₂ emissions are due to reprocessing of the Lubumbashi slag heap: a 14.5×10^{6} 231 tonne hill of mining residue located inside the city, resulting from metallurgical activ-232 ity between 1924 and 1992 (Pe'sa, 2022). Copper, cobalt, and zinc are extracted from the 233 slag heap by La Société Congolaise du Terril de Lubumbashi (2000-present). Reports from 234 the local press attest to the pollution resulting from the operation (Africa Intelligence, 235 2021). Figure 2.e shows annual background-removed TROPOMI-derived NO₂ emissions; because production data is incomplete (STL, 2023; The Carter Center, 2023), emissions and mining-related activity cannot be compared in this case. The most plausible emis-238 sions source in Ndola is the INDENI petroleum refinery plant, located inside the city. 239 Inactive since late 2021, its declining production values (Mwila et al., 2022) match well 240 the TROPOMI-derived NO₂ emissions (R = 0.96, Fig. 2.f). 241

The six Copperbelt NO₂ emission point sources do not coincide with SO₂ enhancements, as shown by the map of mean TROPOMI SO₂ VCD values for the period between January 2019 and July 2022 (Fig. 3). The map shows SO₂ enhancements elsewhere: collocated with the Hwange coal power plant (Zimbawe), the Selous smelter (Zimbawe), and the Chingola and Mufulira smelters (Zambia).

Anthropogenic emissions from the inventory were compared to TROPOMI-derived 247 emissions from our six point sources. Monthly inventory emissions from the following 248 sectors were aggregated: power generation, industrial processes (including mining), road 249 and non-road transportation, residential, fugitive fuel emissions, solvents application and 250 production, and solid waste and wastewater handling. Emissions from the agriculture 251 livestock, agriculture soils, and agriculture waste burning sectors would be part of the 252 background over mines and cities and, thus, were excluded. Aggregated inventory emis-253 sions were converted to NO₂ ($NO_2 = NO_x / 1.32$; Beirle et al. (2019), Dix et al. (2022)). 254 The highest inventory value among the nine data points coinciding spatially with each 255 point source was selected; annual inventory means were calculated and compared to their 256 background-removed TROPOMI-derived emissions counterparts (Table 1 and Fig. S5). 257 The inventory underpredicts mine emissions by 61-96 %, generally overpredicts city emis-258 sions, and does not identify the annual trends in both TROPOMI-derived emissions and 259 mine production. 260

²⁶¹ **5 Discussion**

We have shown that NO_2 from copper/cobalt mining and other industrial activ-262 ities can be identified and emissions quantified from satellite, even in the presence of high 263 background values from biomass burning, soils, and lightening. Furthermore, we have 264 demonstrated strong positive correlations between annual TROPOMI-derived NO₂ emis-265 sions and mine production/refinery throughput. These correlations are mine-dependent 266 and cannot be extrapolated from mine to mine; differences in ore grades and equipment 267 fuel efficiency are probably the main causes. As an example, lower-than-expected 2022 emissions from Lumwana, which reduced R from 1.00 (for 2019-2021) to 0.74 (for 2019-26 2022) (Fig. 2.c), can be traced to a new fleet of trucks and shovels commissioned in 2021 270 (Barrick, 2023) and in operation during 2022 (Lusaka Times, 2022). 271

As a reference, 2019 background-removed TROPOMI-derived NO₂ emissions from Ndola (0.044kg/s, after accounting for point source area) are equivalent to NO₂ emissions from the Miami Fort or the Intermountain coal power plants, both among the ten

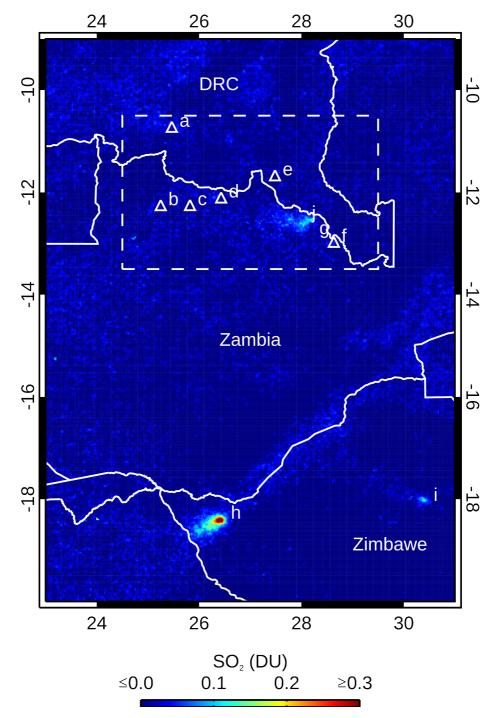


Figure 3. Mean TROPOMI-COBRA SO₂ VCD for January 2019 - July 2022. White triangles show the Copperbelt NO₂ emission point sources analyzed. (a) Kolwezi-Katanga. (b) Sentinel. (c) Lumwana. (d) Kansanshi. (e) Lubumbashi. (f) Ndola. (g) Kitwe. (h) Hwange coal power plant. (i) Selous smelter. (j) Chingola-Mufulira smelters. White lines show country boundaries. Dashed white lines show Copperbelt study area.

largest NO_x emitters in the USA that year (Beirle et al., 2021). Our results demonstrate
that the impact of mining/oil refining on local air quality can be quantified using satellite data; this is particularly important for African regions where mining and other industrial activities proliferate without sufficient air quality monitoring.

We have analyzed the point sources with the highest mean annual TROPOMI-derived emissions in the Copperbelt; our maps do, however, show a string of NO₂ enhancements along the Ndola-Lubumbashi-Kolwezi corridor (Fig. 1) where relatively smaller copper/cobalt mining operations, both industrial and artisanal, exist. Future work should explore the detection limits of this method by analyzing some of these other NO₂ enhancements.

Ground emission measurements were unavailable; thus, we compared backgroundremoved TROPOMI-derived NO₂ emissions to inventory data. Inventory values are lower (higher) for mines (cities) than their TROPOMI counterparts and do not capture the annual trends revealed by TROPOMI. The remoteness of the region, lack of field data, and relatively small size of the point sources may explain at least in part these discrepancies. NO₂ emissions derived from satellite measurements could improve the overall magnitude and temporal (seasonal, inter-annual) trends in inventory emissions. They could also be used to refine regional emission factors, which are not well defined for Africa.

We identified SO₂ enhancements coinciding with smelters and a coal power plant 292 elsewhere, but none colocated with our NO_2 point sources (Fig. 3), despite the fact that some of them either include a smelter (Kansanshi, First Quantum (2023a)) or have one 294 nearby (Lualaba, 40 km southeast of Kolwezi; Ivanhoe Mines (2021)). Lack of SO₂ en-295 hancements indicates use of technologies to convert sulfur oxides released from the ore 296 during smelting into sulfuric acid, a commercial byproduct (Hocking, 2005; Ialongo et 297 al., 2018). The use of such technologies in the Kansanshi and Lualaba smelters has been 298 documented by the mining companies involved (Gray et al., 2020; First Quantum, 2023b; 299 Wang, 2020). 300

301 6 Conclusions

Understanding the environmental effects of high-impact minerals extraction and 302 processing is of great relevance (Hund et al., 2020), particularly if the mining-related ac-303 tivities take place in close proximity to -or even inside of- population centers, as is the 304 case in the Copperbelt region. We have shown for the first time that NO₂ emissions from 305 copper and cobalt mining activities can be identified and measured using TROPOMI satellite data, even in the presence of high background NO₂; this is important in the absence 307 of local air quality monitoring. Furthermore, we have shown, also for the first time, a 308 strong positive correlation between TROPOMI-derived NO₂ emissions and mining/oil 309 refining production data. We note that these correlations are mine-dependent and that 310 changes in the mine's environment (ore grade, fuel efficiency) will affect such correlations, 311 as observed in Lumwana. The lack of SO_2 enhancements colocated with our NO_2 point 312 sources (enhancements identified, though, in power plants and smelters nearby) is con-313 sistent with SO₂ capture and transformation into sulfuric acid, which is then used in mining-314 related processes or commercialized. 315

Because the NO₂ emissions analyzed result from the combustion of fossil fuels by machinery (e.g., trucks, crushers, generators) used extensively in mining operations, these results are relevant to mining in general, regardless of the resource mined. They are also relevant to oil refineries, as shown in Ndola. We hypothesize that our findings apply to fossil-fuel intensive industries in general.

Our results show that NO₂ trend analysis can be used to predict mine production and refinery throughput before companies release their reports or in lieu of these reports in case of non-publicly traded companies, which are not required to publish their activity data. Insufficient emissions from mines claiming high production may be indicative

- of production from a different source. Thus, this method may be useful for improving
- traceability of minerals extracted in conflict areas and smuggled into the global supply
- s27 chain despite existing traceability and tagging schemes, an issue highlighted in the most
- recent releases of the United Nations Yearbook (United Nations, 2019, 2022).

329 **Open Research Section**

Atmospheric data available as follows. TROPOMI NO₂ data (L2_NO2_) are publicly available via the ESA's S5P Pre-Ops interface (<u>https://scihub.copernicus.eu/</u>) using the credentials given there. TROPOMI SO2 COBRA data:

https://distributions.aeronomie.be/?menu=68c9f961bc294141c215e3d64a6ae282#. ERA5 data: https://cds.climate.copernicus.eu/cdsapphttps://cds.climate.copernicus.eu/cdsapp#!/dataset/reanal ysis-era5-single-levels-monthly-means?tab=form (hourly data on single levels) and https://cds.climate.copernicus.eu/cdsapp#!/dataset/reanalysis-era5-pressure-

<u>levels?tab=form</u> (hourly data on pressure levels). CAMS-GLOB-ANT inventory data are publicly available via the https://eccad3.sedoo.fr/ interface following the instructions given there to create a free access account. Mining production data available from mining company reports as follows. First Quantum Minerals Ltd. (Sentinel, Kansanshi):

https://s24.q4cdn.com/821689673/files/doc_downloads/2019-annual-

report/First_Quantum_AR_2019.pdf (2019),

https://s24.q4cdn.com/821689673/files/doc_downloads/2020-annual-

<u>report/First_Quantum_2020_Annual_Report.pdf</u> (2020), <u>https://www.firstquantum-2021-annual-report.com/_files/ugd/acdda3_c51ed134aa184e259d61a629344f98e7.pdf</u> (2021), <u>https://www.first-quantum.com/files/doc_downloads/2022-annual-report/First-Quantum-2022- AR-online.pdf</u> (2022). Barrick Gold Corporation (Lumwana):

https://s25.q4cdn.com/322814910/files/doc_financial/annual_reports/2019/Barrick-Annual_Report-2019.pdf (2019), https://s25.q4cdn.com/322814910/files/doc_financial/annual_reports/2020/Barrick-Annual-Report-2020.pdf (2020),

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https://www.zijinmining.com/upload/file/2020/09/14/9c97a89f8e9c4c59a13f2404f3bb1096.pdf (2019), https://www.zijinmining.com/upload/file/2021/06/09/538a46cc4831452e97b30cae55c9cf97.pdf (2020), https://www.zijinmining.com/upload/file/2022/06/20/771c971d76154257882f58ed03643c07.pdf (2021; no 2022 annual report available at the time of writing). Glencore (Katanga):

https://www.glencore.com/.rest/api/v1/documents/5a08fe1942f92df7f2301ac3681e23aa/glen-2019annual-report-interactive.pdf?download=true (2019),

https://www.glencore.com/.rest/api/v1/documents/3505497f3cb94b24f0c79f5ba32b293b/Glenco re_AR20_Interactive+%281%29.pdf?download=true (2020),

https://www.glencore.com/.rest/api/v1/documents/ce4fec31fc81d6049d076b15db35d45d/GLEN-

2021-annual-report-.pdf?download=true (2021),

https://www.glencore.com/.rest/api/v1/documents/ded10fa92974aa388a43aa9f86f483e9/GLEN-2022-Annual-Report.pdf (2022).

ERA5 data are produced by C3S and CAMS. Contains modified information from C3S and CAMS [2019-2021]. Neither the European Commission nor ECMWF is responsible for any use that may be made of the Copernicus information data contained in this study.

346 Acknowledgments

- ³⁴⁷ SMA thanks William Atkinson for sharing his deep knowledge of minerals and mining;
- Carol Atkinson for her insights and inspiring curiosity; Louisa Emmons for advice re-
- ³⁴⁹ garding models; David Edwards and Bill Randel for helpful NCAR in-house comments.
- ³⁵⁰ This material is based upon work supported by the National Center for Atmospheric Re-
- search, which is a major facility sponsored by the National Science Foundation under
 Cooperative Agreement No. 1852977.

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