

Satellite NO₂ Trends and Hotspots over Offshore Oil and Gas Operations in the Gulf of Mexico

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Key Points:

- Satellite NO₂ records and trends of urban, coastal and deep water areas from 2005 to 2022, are presented
- Classifying NO₂ over the Gulf of Mexico (GOM) under various wind conditions highlights typical patterns in average NO₂ values
- GOM NO₂ hotspots from deepwater platforms were identified by TROPOMI under calm wind conditions, the largest of which is over Mars/Olympus

Abstract

The Outer Continental Shelf of the Gulf of Mexico (GOM) is populated with numerous oil and natural gas (ONG) platforms which produce NO_x (NO_x = NO + NO₂), a major component of air pollution. The Bureau of Ocean Energy Management (BOEM) is mandated to ensure that the air quality of coastal states is not degraded by these emissions. As part of a NASA-BOEM collaboration, we conducted a satellite data-based analysis of nitrogen dioxide (NO₂) patterns and trends in the GOM. Data from the OMI and TROPOMI sensors were used to obtain 18+ year records of tropospheric column (TrC) NO₂ in three GOM regions: 1) Houston urban area, 2) near shore area off the Louisiana coast, and a 3) deepwater area off the Louisiana coast. The 2004-2022 time series show a decreasing trend for the urban (-0.027 DU/decade) and near shore (-0.0022 DU/decade) areas, and an increasing trend (0.0019 DU/decade) for the deepwater area. MERRA-2 wind and TROPOMI NO₂ data were used to reveal several NO₂ hotspots (up to 25% above background values) under calm wind conditions near individual platforms. The NO₂ signals from these deepwater platforms and the high density of shallow water platforms closer to shore were confirmed by TrC NO₂ anomalies of up to 10%, taking into account the monthly TrC NO₂ climatology over the GOM. The results presented in this study establish a baseline for future estimates of emissions from the ONG hotspots and provide a methodology for analyzing NO₂ measurements from the new geostationary TEMPO instrument.

Plain Language Summary

Oil and natural gas operations emit nitrogen oxides (NO_x), which are major air pollutants and precursors to ground-level ozone. The Bureau of Ocean Energy Management (BOEM) agency is responsible for managing planned oil and natural gas (ONG) activity on the outer continental

45 shelf, and is mandated to ensure related emissions do not degrade air quality of coastal states. In
46 collaboration with BOEM, we used satellite data from the OMI and TROPOMI sensors to
47 construct an 18+ year record of tropospheric nitrogen dioxide (NO₂), a proxy for NO_x, in the
48 Gulf Coast region. These time series focused on three areas: 1) Houston urban, 2) off the
49 Louisiana coast, and 3) deepwater Gulf off Louisiana. These regions experienced changes in
50 tropospheric column NO₂ of -13.7%, -5.8% and +5.4% per decade, respectively. We also
51 identified NO₂ hotspots from ONG platforms using TROPOMI NO₂ averages under calm wind
52 conditions. The ONG deepwater platforms enhance NO₂ background amounts by 7-13% on
53 average, and up to 25% for the Mars and Olympus platforms combined. The results presented
54 here will facilitate our work on emissions estimates from these sources and on applications to the
55 recently launched TEMPO instrument.

56

57 **1. Introduction**

58

59 Nitrogen dioxide (NO₂), a component of NO_x (NO_x = NO + NO₂) and classified as a criterion
60 pollutant by the Environmental Protection Agency (EPA), is produced from fuel combustion.
61 Anthropogenic sources of NO₂ include fires, vehicular emissions, power plants and other
62 industrial activities such as oil and natural gas (ONG) production. In large quantities, NO₂ causes
63 respiratory problems from prolonged exposure. Furthermore, NO₂ is a major precursor to
64 tropospheric ozone (O₃), another criteria pollutant responsible for damaging effects on lungs and
65 premature mortality (Bell et al., 2006). Amounts of NO₂ are measured with in-situ analyzers,
66 typically reporting in mixing ratio, or through remote sensing instruments that report column
67 amounts. Ground-based remote sensors for total column NO₂ (TC NO₂, Piters et al., 2012)
68 include the Pandora spectrometer (Herman et al., 2009). Airborne remote sensors, e.g., the Geo-
69 CAPE Airborne Simulator (GCAS) (Nowlan et al., 2018; Judd et al., 2019) and the
70 Geostationary Trace gas and Aerosol Sensor Optimization (GEO-TASO) (Nowlan et al., 2016)
71 measure the NO₂ column amount below the aircraft. From space, TC NO₂ is measured with
72 satellite ultraviolet-visible (UV-Vis) sensors. A long TC NO₂ record exists thanks to a series of
73 satellite sensors: the Global Ozone Monitoring Experiment (GOME, Burrows et al., 1999;
74 Richter et al., 2002) and GOME-2 (Richter et al., 2011; Munro et al., 2016) instruments, the
75 Ozone Monitoring Instrument (OMI) (Levelt et al., 2006; Levelt et al., 2018), and the
76 Tropospheric Monitoring Instrument (TROPOMI) instrument launched in 2017 (Van Geffen et
77 al., 2012). TROPOMI also supplies a tropospheric column (TrC) NO₂ product, which has been
78 used for a range of air quality applications.

79 Satellite instruments are very useful for providing column NO₂ data over areas without
80 surface monitoring, especially over water. Over the past decade a number of studies have
81 compared satellite TC NO₂ with both in-situ and remotely sensed NO₂ in coastal areas. Land-sea
82 interactions, e.g., sea-breeze and other dynamical factors, show how challenging satellite NO₂
83 validation can be over both sides of the land-water interface. The Korea-United States Ocean
84 Color experiment (KORUS-OC) (Tzortziou et al., 2018; Thompson et al., 2019), around the
85 Korean Peninsula in 2016, found that complex interactions of advected pollution and
86 meteorology determined whether satellite TC NO₂ column amounts correlated with shipboard
87 Pandora TC NO₂ and in-situ NO₂ measurements. Similarly, the Ozone Water-Land
88 Environmental Transition Study (OWLETS) projects over the southern Chesapeake Bay region
89 in 2017 (Sullivan et al., 2018; Dacic et al., 2020) and near Baltimore in 2018 (Sullivan et al.,
90 2020; Kotsakis et al., 2022) discovered that the accuracy of satellite TC NO₂ data depended on
91 resolution (pixel size), cloud-cover, pollution amount and whether the satellite was measuring

92 over land or water. Other campaigns with TC NO₂ measurements in coastal areas include
93 DISCOVER-AQ in Baltimore (2011) (Tzortiou et al., 2013; Reed et al., 2015) and Houston
94 (2013) (Judd et al., 2019; Choi et al., 2020), and the Deposition of Atmospheric Nitrogen to
95 Coastal Ecosystems (DANCE) campaign in (Martins et al., 2016; Kollonige et al., 2019).

96 In the Gulf of Mexico (GOM), a notable source of NO_x is from ONG exploration and
97 production sites. The above-mentioned campaigns, while investigating air quality in coastal
98 regions, did not focus on areas of concentrated offshore ONG activity or validation of satellite
99 NO₂ near ONG sources. We addressed these issues in a 3-year study that NASA undertook in
100 collaboration with the Bureau of Ocean Energy Management (BOEM, Department of Interior).
101 BOEM is the agency responsible for managing ONG exploration, development, and production
102 plans in the U.S. Outer Continental Shelf (OCS). The agency specifically has air quality
103 jurisdiction for OCS emissions from ONG exploration and development to the west of 87.5°. It is
104 also mandated to ensure that criteria pollutant emissions from these activities are in compliance
105 with the national ambient air standards to the extent that the activities do not significantly affect
106 the air quality of any state. BOEM tracks industry-reported NO_x emissions from ONG operations
107 in monthly inventories (Wilson et al., 2018). However, due to lack of air quality monitoring on
108 the OCS, the reported emissions remain unvalidated. NASA and BOEM carried out a feasibility
109 study from 2017 to 2020 to determine whether satellite data could be used to monitor NO₂ over
110 the GOM and discriminate regional sources and/or resolve pollution from individual platforms.
111 Preliminary results were summarized in two documents by Duncan (2020) and Thompson
112 (2020). These were followed by detailed reports on a 2019 field campaign (Satellite Coastal and
113 Oceanic Atmospheric Pollution Experiment), SCOAPE-I (Thompson, 2020; Thompson et al.,
114 2023), along the Louisiana coast.

115 The SCOAPE-I cruise took place 10-18 May 2019 aboard the *Research Vessel Point Sur*.
116 One of the goals of this campaign was to measure in-situ NO₂ levels along the Louisiana coast
117 with a cruise track designed to sample smaller near-shore ONG operations, over open water, and
118 near large deepwater ONG platforms farther away from the coast. The deepwater platforms
119 primarily produce oil and flare excess gas; thus, they usually have larger individual platform
120 NO_x emissions. For NO₂, the *Point Sur* was equipped with a NO₂ in-situ analyzer and a Pandora
121 spectrometer for measuring TC NO₂ amounts. Pandora measurements were taken during daytime
122 in cloud-free conditions. Measurements of NO₂ were also collected with Pandora, satellites, and
123 an NO₂ analyzer at the Louisiana Universities Marine Consortium (LUMCON; Cocodrie, LA;
124 29.26°, 90.66°) SCOAPE-I port during the cruise and the three weeks prior.

125 During the SCOAPE I cruise, satellite (OMI, TROPOMI) and the shipboard Pandora total
126 column (TC) NO₂ levels were elevated in the vicinity of ONG platforms as confirmed by
127 numerous coincident NO₂ spikes from the shipboard analyzer. However, neither the satellite nor
128 Pandora TC NO₂ responses to emissions were as large as surface NO₂ increases. Comparisons
129 between NO₂ column amounts from satellite and surface Pandoras showed good agreement
130 during SCOAPE I - within 13% over water and 5% over land in clear sky conditions - and NO₂
131 signals from selected ONG platforms could sometimes be isolated (Thompson et al., 2023).
132 However, consistent quantification of NO₂ sources was not possible due to cloud cover, satellite
133 sampling frequency (one overpass daily) and relatively coarse spatial resolution compared to
134 platform size, factors all amplified by the short duration of the cruise. Two air quality regimes,
135 differentiated by prevailing wind direction, were characterized by surface and satellite
136 measurements during SCOAPE I: clean marine air over deepwater (onshore wind from remote
137 marine locations) and polluted continental air near shore (wind from land). In between, elevated
138 NO₂ near-shore can result from nearby pollution, from deepwater regions *or* from the continent.

139 There is now nearly five years of TROPOMI and 18+ years of OMI observations, both
140 TC NO₂ and TrC NO₂. This prompts us to conduct a more comprehensive study of satellite NO₂
141 over the GOM, examining regional and temporal variability. We use OMI and TROPOMI TrC
142 NO₂ data to: 1) analyze the long-term NO₂ record over three prototype regions within the GOM
143 and 2) identify NO₂ hotspots near ONG operations using wind-classified TROPOMI data. The
144 longer time-series are used to determine trends in NO₂. Anomalous calm-wind TROPOMI data
145 pinpoint major and lesser-emitting platforms. The results, summarized in Section 3.1 and 3.2,
146 provide a baseline for the longer-term goal of monitoring GOM ONG NO_x emissions.
147 Identifying the hotspots is crucial for BOEM’s mission and demonstrates the ability to monitor
148 ONG pollution with remote sensing instruments in anticipation of the geostationary
149 Tropospheric Emissions: Monitoring of Pollution (TEMPO) data. The results for this second
150 objective are found in Section 3.3.

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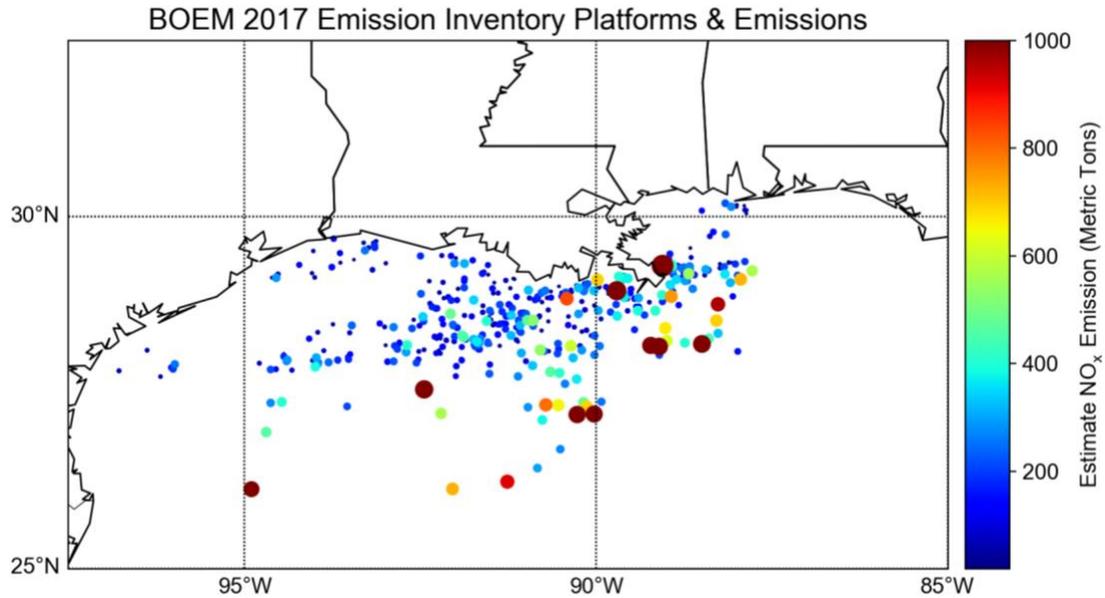
152 **2. Methodology**

153

154 **2.1 Description of study domain**

155 The focus area is the Outer Continental Shelf (OCS) of the Gulf of Mexico (GOM), an area
156 populated with numerous oil and natural gas platform operations (Figure 1). There are two main
157 types of platforms: shallow water and deepwater platforms. Deepwater platforms are further
158 from the coast, more isolated, and produce NO_x emissions due to gas flaring. Shallow water
159 platforms generally produce less NO_x than deepwater platforms but are greater in density near
160 the shore. The majority of the platforms in the GOM produce less than 1000 metric tons of NO₂
161 per year with exception of the deepwater platforms away from shore, according to the 2017
162 BOEM inventory (Figure 1; Wilson et al., 2018). The domain of this study is between 25°N and
163 30° N latitude and 95° W and 87.5° W longitude. This includes most of the OCS of the GOM
164 over which BOEM has jurisdiction (200 nautical miles beyond state jurisdictions). We also
165 consider the Houston, TX, urban area as a reference in comparison to the areas over water. In
166 particular, Houston was chosen because it is the highest NO_x emitting region along the Gulf
167 coast and has been the location of multiple air quality campaigns such as DISCOVER-AQ (Judd
168 et al., 2019; Nowlan et al., 2018; Choi et al., 2020) and TRACER-AQ in 2021 (Jensen et al.,
169 2022; Judd et al., 2021).

170



171 **Figure 1:** Locations of the Gulf of Mexico ONG platforms in the BOEM 2017 emission
 172 inventory. Larger dots and corresponding colors indicate the platforms with the highest annual
 173 NO_x emissions.
 174
 175

176 The study domain was further divided into smaller regions to compare areas that are expected
 177 to have contrasting NO_x emissions and observed NO_2 amounts. A deepwater area, a near shore
 178 area and an urban area were defined and shown as a green, red and orange box in Figure 2,
 179 respectively. The near shore area covers parts of both BOEM (federal) and Louisiana state
 180 jurisdictions, and includes numerous shallow water platforms within about 100 km from the
 181 Louisiana coast. The latitude bound of the near shore area is 28.3°N and 29.3°N in this analysis.
 182 The defined deepwater area is between 27° and 28.3°N and includes several deepwater
 183 operations with NO_x emissions greater than 500 metric tons. In this study, the deepwater area can
 184 also be treated as being close to background NO_2 levels, since marine air is clean and the
 185 deepwater platforms are relatively isolated.
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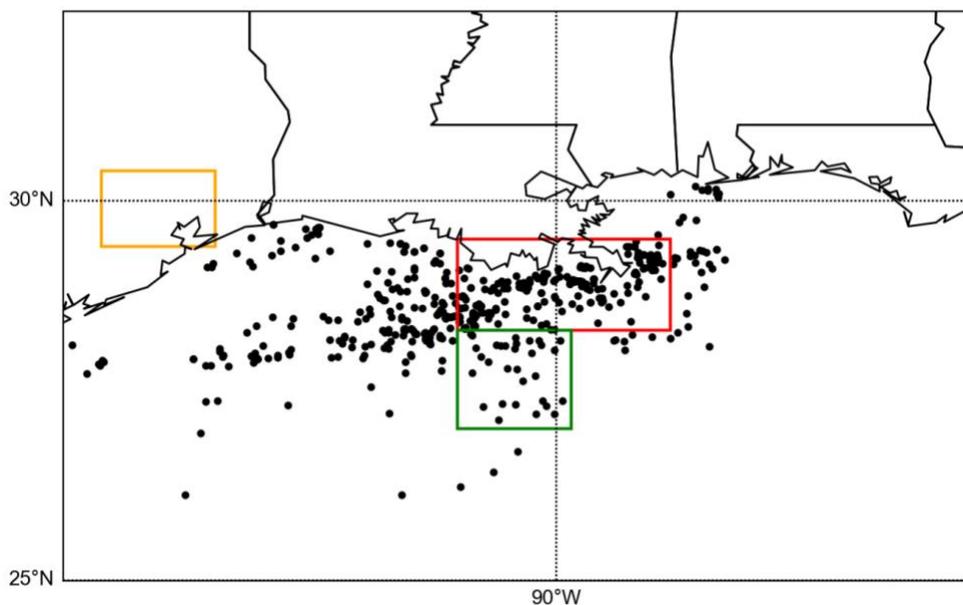


Figure 2: Map of the three areas of focus for which OMI and TROPOMI time series are calculated: Shallow water off the east Louisiana coast (red box), GOM deepwater (green box) and Houston, TX, metropolitan area (orange box). Each black dot represents a platform or facility in the BOEM OCS 2017 emissions inventory.

2.2 Datasets

For this work, we used satellite datasets from the Ozone Monitoring Instrument (OMI) (Levelt et al., 2006; Levelt et al., 2018) and TROPOMI (Veefkind et al., 2012). Located onboard NASA’s polar orbiting Aura satellite, OMI was launched in 2004 and its data record began in October of that year. OMI collects observations over a particular location about once a day at a spatial resolution of $13 \times 24 \text{ km}^2$ at nadir and $24 \times 160 \text{ km}^2$ at the edge of the swath. The satellite is sun synchronous and makes an overpass at around 1300-1400 local time. In this study we use the high-resolution OMI Tropospheric NO_2 Version 4 dataset (Lamsal et al., 2021; https://avdc.gsfc.nasa.gov/pub/data/satellite/Aura/OMI/V03/L3/OMNO2d_HR), which contains several improvements to air mass factors (AMFs) compared to Version 3. In particular, this version incorporates improved cloud algorithms, a geometric Lambertian Equivalent Reflectance (GLER) product and improved terrain pressure calculations into the NO_2 retrieval. This Level 3 (L3) gridded research product has a resolution of $0.1^\circ \times 0.1^\circ$ – an increase from the $0.25^\circ \times 0.25^\circ$ of the original Level 3 dataset. At times the spatial coverage of OMI is impacted by the row anomaly (Torres et al., 2018), a physical instrument issue which obstructs some of the instrument’s field of view and therefore affects radiance measurements.

The TROPOMI instrument was launched by the European Space Agency on the European Union’s Copernicus Sentinel 5 Precursor (S5P) satellite in October 2017, with the data record beginning May 2018. The overpass of TROPOMI occurs in early afternoon, within about 0.5 hr. of OMI. The resolution of the instrument is currently $3.5 \times 5.6 \text{ km}^2$ at nadir ($3.5 \times 7 \text{ km}^2$ prior to August 2019). Like other polar orbiting instruments, TROPOMI provides daily global coverage, although only about once per day at any given location. TROPOMI’s NO_2 algorithms use a differential optical absorption spectroscopy (DOAS) technique on radiances in the 405–465 nm spectral window. The spectral radiances are converted into slant column

217 densities (SCD) of NO₂ between the instrument and the Earth's surface (van Geffen et al., 2020).
218 AMFs are then used to convert the slant column into a vertical column density (VCD). For
219 obtaining the tropospheric NO₂ column, the stratospheric portion is subtracted from the total
220 SCD using global model estimates (Boersma et al., 2004; Boersma et al., 2007). The algorithms
221 have been updated throughout the course of TROPOMI's operation, resulting in multiple
222 versions of the data. The research dataset S5P-PAL ([https://data-portal.s5p-
223 pal.com/products/no2](https://data-portal.s5p-pal.com/products/no2)) was developed to apply the new algorithm (v2.3) to the older radiances,
224 essentially homogenizing the data with respect to retrieval differences.

225 Lastly, we use the Modern Era Retrospective Analysis for Research and Applications
226 Version 2 (MERRA-2, Gelaro et al., 2017) for wind analysis incorporated into calculating the
227 satellite NO₂ time series. MERRA-2 is derived from the GEOS-5 data assimilation system and
228 contains meteorological variables on a 0.5° × 0.625° grid for 42 standard pressure levels. The
229 variables used in the analysis are the U and V components of the wind which are used to derive
230 vector wind speed and direction.

231

232 **2.3 Satellite NO₂ Time Series**

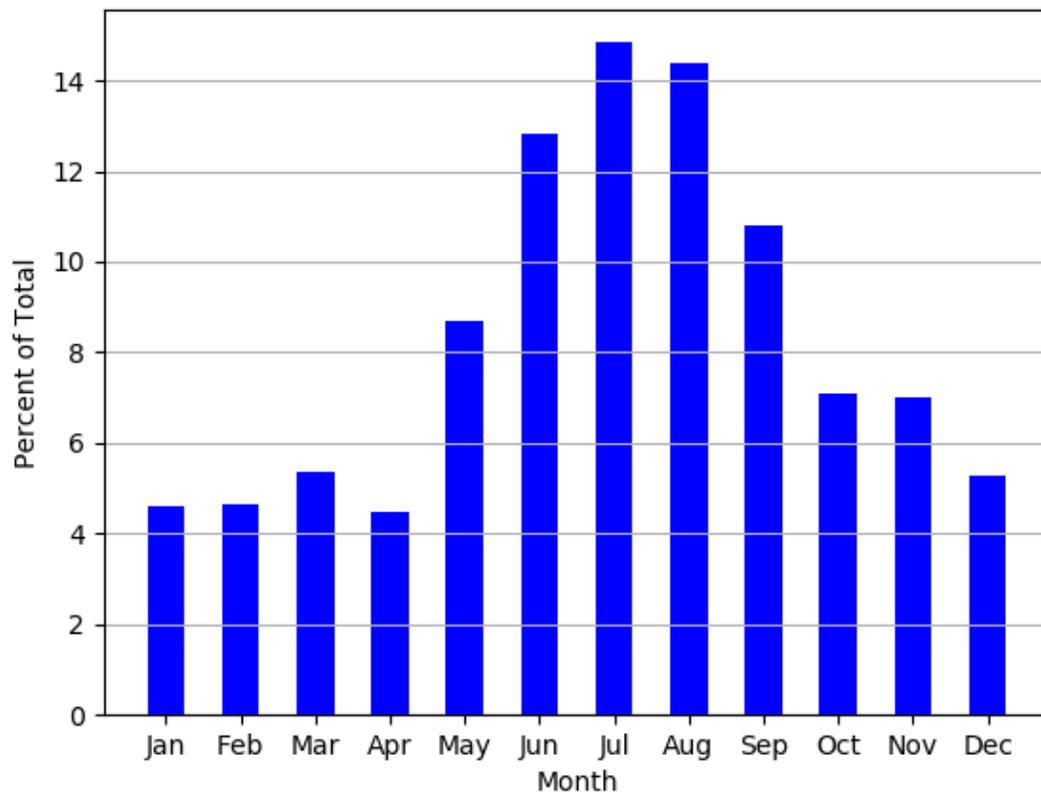
233 The time series in this work consist of monthly averages of TrC NO₂. For OMI, a
234 monthly version of the L3 high-resolution dataset is already available as a research product. The
235 TROPOMI data were compiled by finding all overpasses over our study region in that month, re-
236 gridding each daily file to a 0.01° × 0.01° grid and calculating the average for each grid point.
237 The recommended quality assurance (QA) value of 0.75 was used as a threshold for filtering bad
238 quality pixels. Next, the grid cells in each area previously described (deepwater, near shore and
239 urban) were averaged for each month to obtain the time series for that particular region. From the
240 resulting time series of monthly averages we also calculated the 12 month moving averages to
241 account for the seasonality in the NO₂ time series. A trend line was fitted to the moving average
242 to obtain an overall linear trend over the full record. The trends are presented in Section 3 along
243 with the 95% confidence intervals.

244 Since meteorological regimes drove much of the variability in TC NO₂ in coastal areas
245 during SCOAPE-I (Thompson et al., 2023), we also analyzed how the time series differ
246 according to different wind speed and directions (source regions). For this objective, daily
247 MERRA-2 wind data from 2005-2022 were used to restrict the NO₂ averaging to days based on
248 three cases: 1) Wind from the north (land) at greater than 10 ms⁻¹ 2) wind from the south (GOM)
249 at greater than 10 ms⁻¹ and 3) calm winds of less than 5 ms⁻¹. The 10 ms⁻¹ threshold was chosen
250 to ensure that sufficient transport was occurring that day within the lower levels of the
251 atmosphere. For the north and south wind conditions, we defined the degree bounds as 120° to
252 240° and 300° to 60°, respectively. Note that here we use meteorological wind directions where
253 due north is 360°. For the calm wind case, all wind directions are considered. It is important to
254 note that 140-160 days out of the year are ignored, such as those with wind speeds between 5 and
255 10 ms⁻¹. For the Houston, TX box, the wind direction bounds were rotated by 45°
256 counterclockwise in order to account for orientation of land and sea with respect to the city. The
257 overall objective of this analysis was to determine how much the NO₂ column amount averages
258 differ based on land vs. marine source regions, as well as cases with calm conditions and less
259 regional transport. The MERRA-2 winds were evaluated for all MERRA-2 grid points within
260 each box in Figure 2. A specific day was categorized if the wind direction was within the degree
261 bounds and wind speed condition was met at all points. The 950 hPa pressure level was used
262 because we are generally interested in the wind in the boundary layer but not specifically at the
263 surface in the case that there is transport occurring aloft. The model surface winds also tend to

264 carry more uncertainty than at levels aloft. To best coincide with the overpass of OMI and
265 TROPOMI, we only used the wind information at 18 UTC (12-1 pm local time).

266 Once the sets of days corresponding to each wind criterion were compiled, 3-month
267 averages were computed from those days for each NO₂ time series. Three-month periods were
268 used to account for sample size issues; some months have too few days of a wind criterion being
269 met. For example, calm winds are more common in the summer than winter according to the
270 climatology compiled from the MERRA-2 data (Figure 3). Aside from using the selected days
271 for each case, the procedure for averaging the TROPOMI Level 2 (L2) and OMI L3 gridded data
272 files was the same. This analysis yielded three time series for each area, corresponding to the
273 three wind criteria. These results and their implications are discussed in Section 3.
274

Distribution of Calm Wind Days (< 5 m/s) by Month (2005-2022)



275
276 **Figure 3:** Distribution of number of days for each month for which the MERRA-2 950 hPa wind
277 evaluated in the near shore area was less than 5 ms⁻¹ (calm wind case). It is expressed as a
278 percentage of the total number of days over the 18 years of the OMI record (2005 through 2022).
279

280 2.4 TROPOMI NO₂ Averages and Anomalies

281 The wind-based averaging was extended to TROPOMI data, with the goal of identifying
282 NO₂ hotspots. This was done on an annual basis by calculating an average of all days in each
283 year that fit the calm wind case (winds < 5 ms⁻¹). The maps with average TrC NO₂ are shown
284 and described in Section 3.3. The same quality assurance threshold (0.75) and re-gridding
285 technique was used as for the complete TROPOMI time series. To account for seasonality and
286 differences in NO₂ between months, we also calculated TROPOMI NO₂ anomalies for 2018-
287 2022. The first step was computing a climatology for every month by averaging all days during

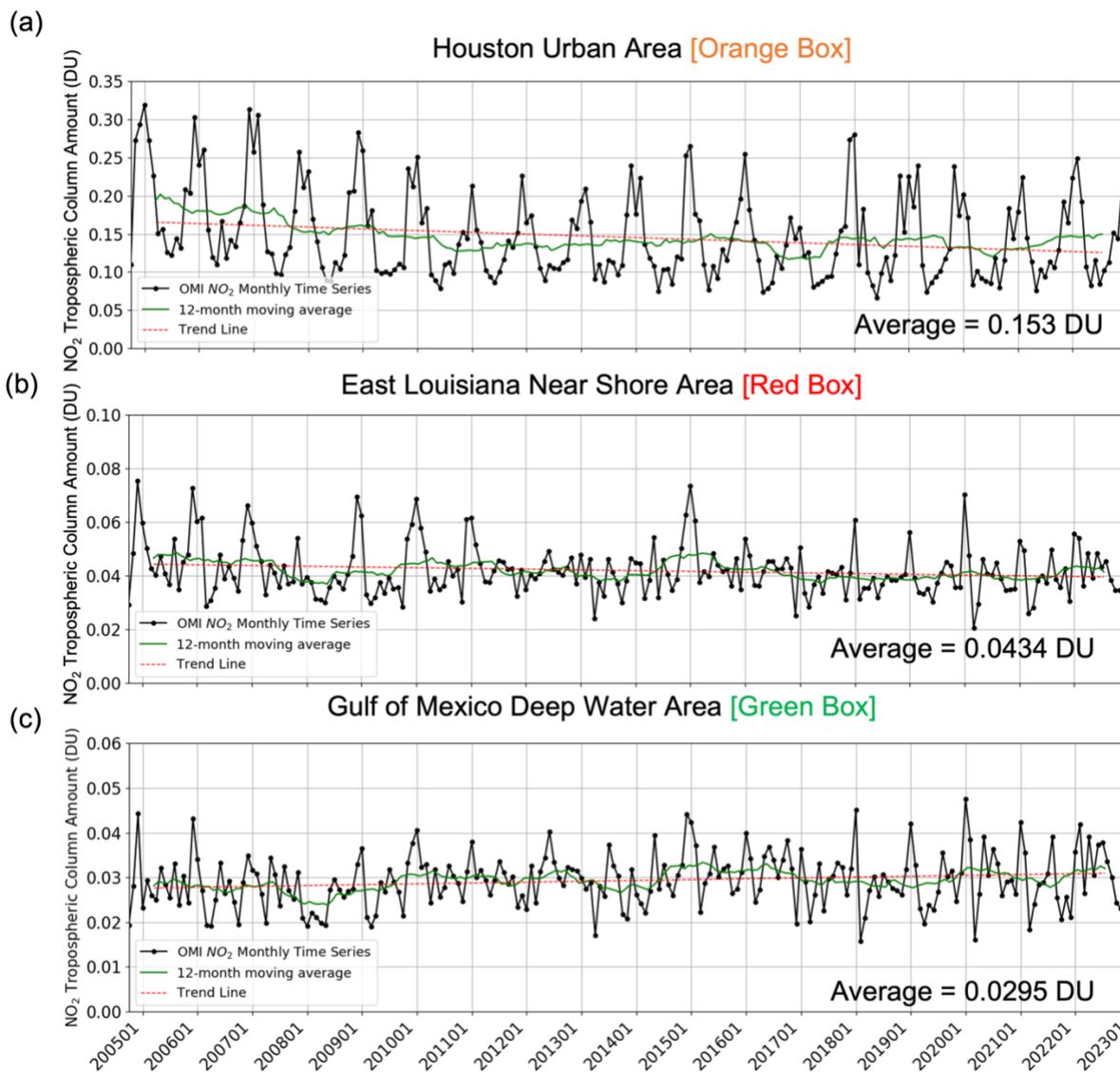
288 the TROPOMI record for each individual month. Next, we separated the calm wind days by
289 month and for each calculated the percent TrC NO₂ difference between the individual day and
290 the climatology for the same month. Over the roughly 4.5 years of TROPOMI's data record,
291 there were 450 individual calm wind day NO₂ anomalies calculated. An average was taken of
292 this set of anomalies to obtain a single gridded anomaly (see Section 3.3) that describes the
293 enhancement or reduction of NO₂ over each grid cell that also accounts for seasonal changes in
294 NO₂ amounts.

295 **3. Results**

296

297 **3.1 Satellite NO₂ Time Series**

298 Figure 4 shows the 2004-2022 OMI time series for the three boxes defined in Figure 2.
299 The red dashed trend lines were calculated from the 12-month moving averages to remove NO₂
300 seasonality. In the Houston, TX area (Figure 3a), the time series exhibits large seasonal
301 fluctuations, in most cases over a factor of two from the winter to summer months. This is
302 typically due to the differences in NO₂ lifetime in winter and summer months. The lifetime
303 varies from 2 to 5hr during the daytime in summer (Beirle et al., 2011) and 12–24 hr during
304 winter (Shah et al., 2020). The amplitudes of the peaks are noticeably higher in the first four
305 years (2005-2009) of the time series compared to the most recent decade. There is an overall
306 negative trend of -0.027 ± 0.0055 DU per decade with 13.7% decrease per decade, much of it due
307 to the reduction of NO₂ in the first 5 years of the time series. Similar trends are also observed in
308 urban areas throughout the U.S (Lamsal et al., 2015; Krotkov et al., 2016, Goldberg et al., 2021).
309 After 2010, the NO₂ remains relatively constant. Over the entire time series the average value is
310 0.153 DU and this average value is closer to the minima of the NO₂ annual cycles due to the
311 troughs of the annual cycle lasting several months, as opposed to 1-3 month peaks in winter.
312



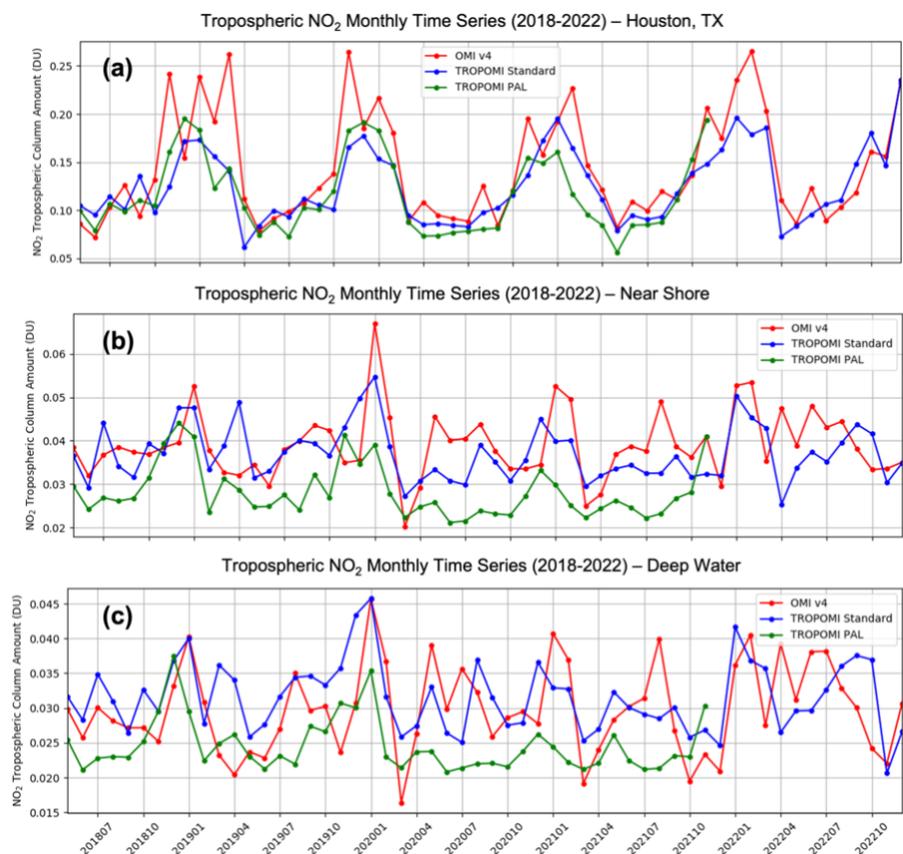
313
 314 **Figure 4:** Time series of OMI TrC NO₂ monthly averages for the boxes shown in Figure 2,
 315 between late 2004 and 2022: (a) Houston, TX urban area (orange on Figure 2), (b) Near-shore
 316 (red on Figure 2), (c) GOM deepwater area in the GOM (green on Figure 2). The 12-month
 317 moving average (green line) and the linear trend line (dashed red) over the time series are also
 318 plotted.
 319

320 Unlike the urban area, in the near shore area (Figure 4b) there is a less-defined NO₂
 321 seasonality at the coast and over water. Although peaks and troughs in column NO₂ exist in
 322 2005-2010, thereafter the time series becomes highly variable from month to month. The trend in
 323 this region was -0.0022 ± 0.0008 DU per decade (5.8% decrease per decade), and while still
 324 negative, is a lower magnitude than that of Houston by a factor of 10. The negative trend
 325 indicates the influence of relatively polluted land areas to the north, such as New Orleans;
 326 however, with fewer significant local sources, the potential for trends resulting from NO₂
 327 emissions reductions is lower over-water. The average value for the near shore region is 0.0434
 328 DU, about 28% of the urban Houston area value. It is important to note that while OMI can be

329 useful for remote sensing over water (e.g., Thompson et al., 2023), the data tend to be noisy on a
330 day to day basis.

331 The deepwater area is characterized by a noisy time series with no discernable seasonal
332 pattern (Figure 4c). Since all pixels in the box are used to calculate the average column amount
333 value for each month, the influences of deepwater ONG operations in this area, which are
334 relatively small compared to the pixel size, are likely washed out. In Section 3.3 we also show
335 time series for NO₂ hotspots over individual ONG platforms without including the rest of the
336 deepwater area. The overall average value in the deepwater area was 0.0295 DU, around 67%
337 that of the near shore area and 19% of Houston. There is a slight increase of 5.4% per decade in
338 this area with a positive trend (0.00189 ± 0.00054 DU per decade). The positive trend may result
339 solely from noise due to the low NO₂ column amounts. However, we also note that there was an
340 increase in deepwater ONG operations in the last decade which could have contributed to this
341 trend (Section 3.3). Only the Houston trend is statistically significant given that the 95%
342 confidence interval indicates an error uncertainty of 5.5%. For the near shore and deepwater
343 areas, the uncertainty is around 30% for each. Given the higher uncertainty and smaller trend
344 values, we cannot make a conclusive determination on whether the ONG activity drives these
345 trends.

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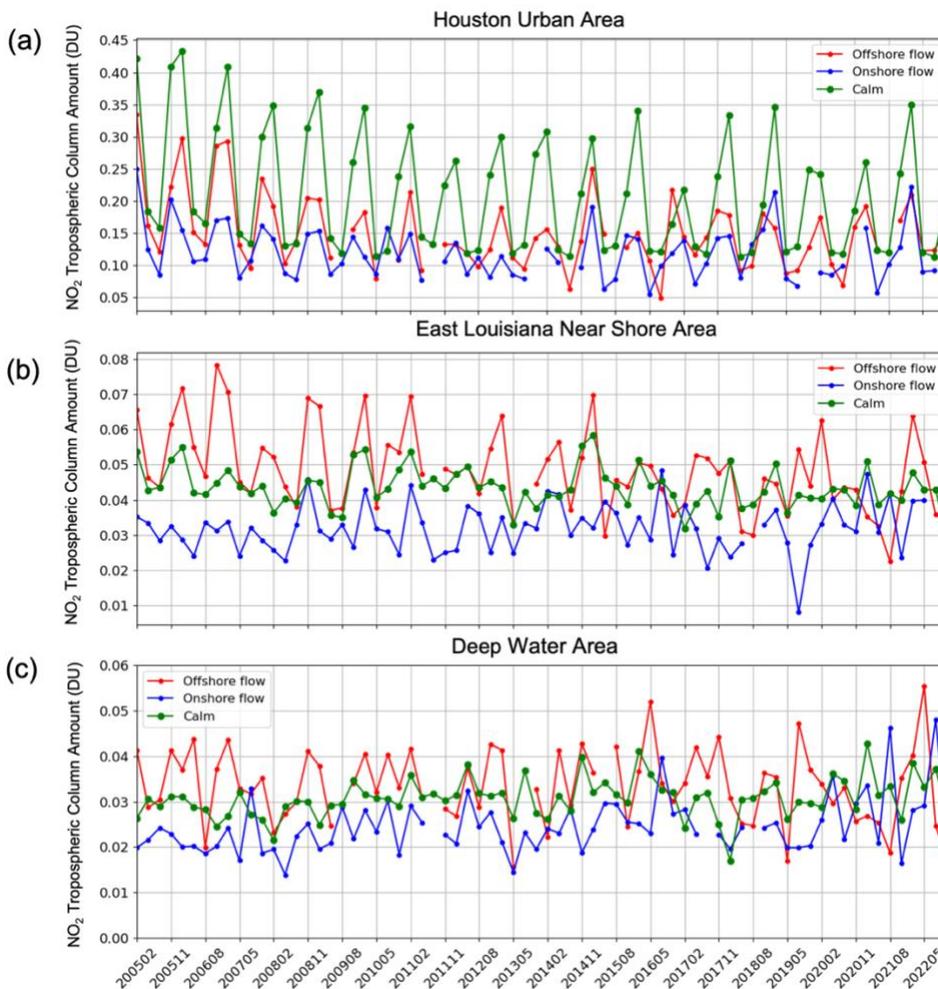


347 **Figure 5:** Comparisons of the 2018-2022 TrC NO₂ time series from OMI v4, the standard
348 TROPOMI product and the reprocessed TROPOMI PAL product for a) Houston, TX urban area,
349 (b) Near-shore, (c) GOM deepwater area in the GOM. Note that the TROPOMI PAL dataset is
350 currently only available through 2021.

351

352

353 We calculated the NO₂ time series from all three satellite datasets over our three regions
 354 to provide context for the values described above. The time series of OMI, standard TROPOMI,
 355 and the TROPOMI PAL datasets (Figure 5) from 2018-2022 show how the observations differ.
 356 OMI averages 7.5% and 4.5% higher than TROPOMI in the near shore (Figure 5b) and
 357 deepwater areas (Figure 5c), respectively. This makes sense given that in the urban area OMI is
 358 clearly higher than TROPOMI observations on average (Figure 5a). These time series also show
 359 that the standard TROPOMI product is consistently higher than the PAL by around 21.5% in the
 360 deepwater box and 15.7% in the near shore box. The difference between the PAL and standard
 361 TROPOMI dataset was not found to be large over a polluted area like Houston, with the standard
 362 product being on average within 10% of the PAL between mid-2018 and late 2021.



363 **Figure 6:** Time series of OMI TrC NO₂ seasonal averages based on wind condition, for: (a)
 364 Houston, TX urban area, (b) Near-shore, and (c) GOM deepwater area. The three conditions for
 365 (b) and (c) are: MERRA-2 950 hPa winds > 10 ms⁻¹ from the south quadrant (blue line), winds >
 366 10 ms⁻¹ from the north quadrant (red line) and winds less than 5 ms⁻¹ from any direction (green
 367 line). For (a), the directions were adjusted to southwest (blue) and northwest (red) quadrants
 368 respectively.
 369

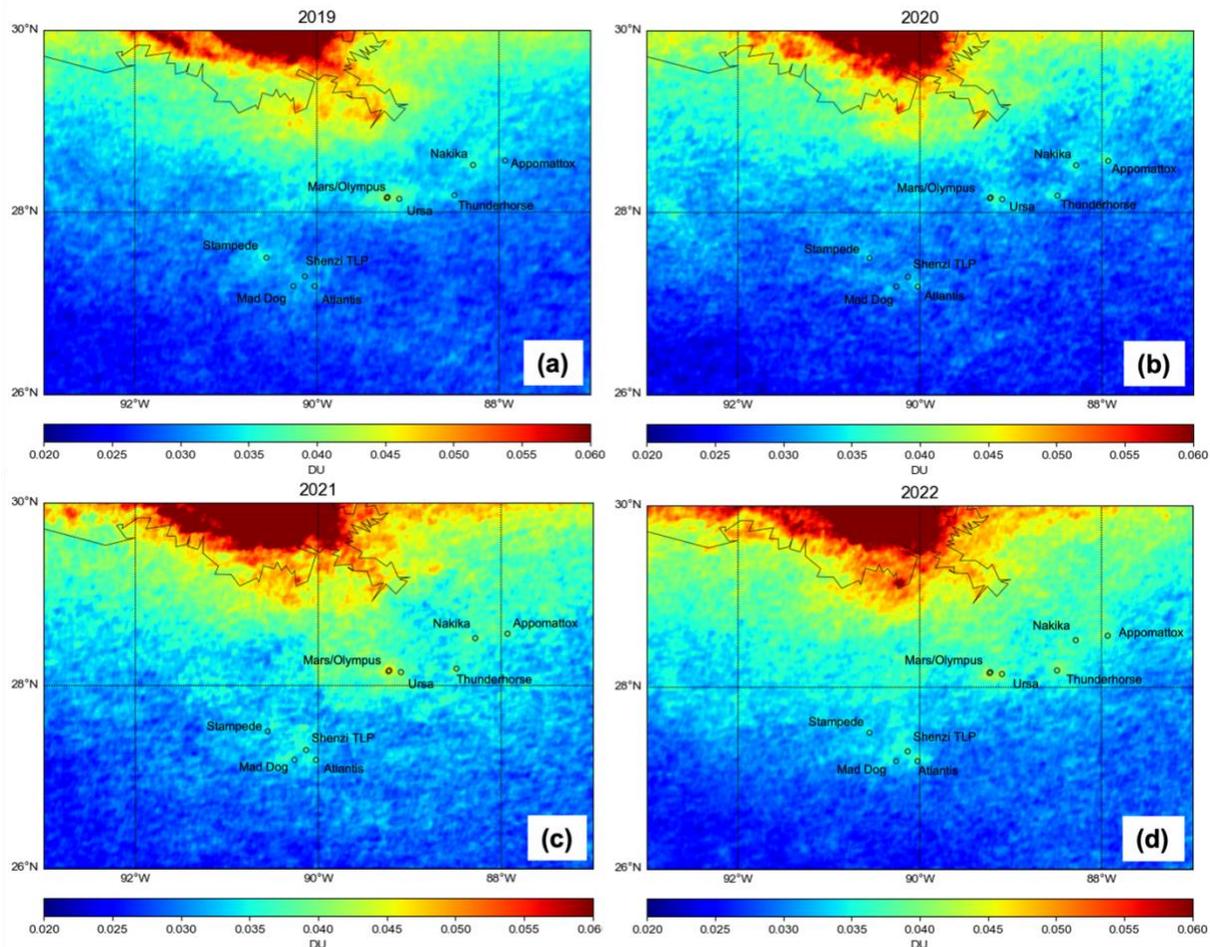
370
 371 **3.2 Wind-based Time Series**

372 The time series, with the influence of different wind conditions described in Section 2,
373 are displayed in Figure 6. Missing points are the 3-month time periods when there were not
374 enough days to meet the wind conditions (at least 10). For the Houston area (Figure 6a), calm
375 winds clearly result in the highest average NO₂ column amount (0.207 DU) because there is little
376 transport of emissions away from the city. This is about a 35% increase above the average of the
377 entire Houston time series (Figure 4a). Offshore flow (wind from other land areas toward the
378 GOM) produces an average monthly column amount of 0.151 DU, while onshore flow results in
379 an average value of 0.118 DU. This indicates that for the most part offshore sources do not
380 impact onshore areas given their small magnitude of NO_x emission. As expected, in the near
381 shore and deepwater areas (Figure 6b and 6c), the calm wind time series fall in between the
382 offshore and onshore wind time series. The trends of the calm wind case in near shore and
383 deepwater areas are -0.0027 DU per decade and 0.0021 DU per decade respectively. Although
384 these NO₂ amounts and trends are relatively small, this equates to a total increase of 13.5% for
385 the deepwater area, and 8.5% decrease for the near shore area over the OMI record. The NO₂
386 trends corresponding to calm wind conditions can also better describe trends over ONG
387 operations since we eliminate days with significant transport of clean or polluted air masses.
388 Average TrC NO₂ amount for the onshore flow case in the deepwater was 0.024 DU which can
389 be considered very close to a typical background value over clean marine areas. For the calm
390 wind case the average value was 0.032 DU, only slightly lower than the offshore case (0.0335
391 DU). The small difference between the two can be partially explained by the significant dip in
392 2020 for the offshore flow case, most likely due to the COVID-19 lockdowns. Table 1 summas
393 the average column amounts for the original time series and wind-based time series. For the near
394 shore and deepwater areas, the calm wind time series averages were close to the overall average.
395
396 **Table 1:** Average OMI Column TrC NO₂ for the wind-based time series and the original time
397 series.

Time Series NO₂ Tropospheric Column Amount Averages (DU)

| Wind condition | Urban | Near shore | Deepwater |
|-----------------|-------|------------|-----------|
| Offshore flow | 0.151 | 0.0495 | 0.0335 |
| Onshore flow | 0.118 | 0.0307 | 0.0231 |
| Calm | 0.209 | 0.0440 | 0.0307 |
| None (all days) | 0.153 | 0.0434 | 0.0295 |

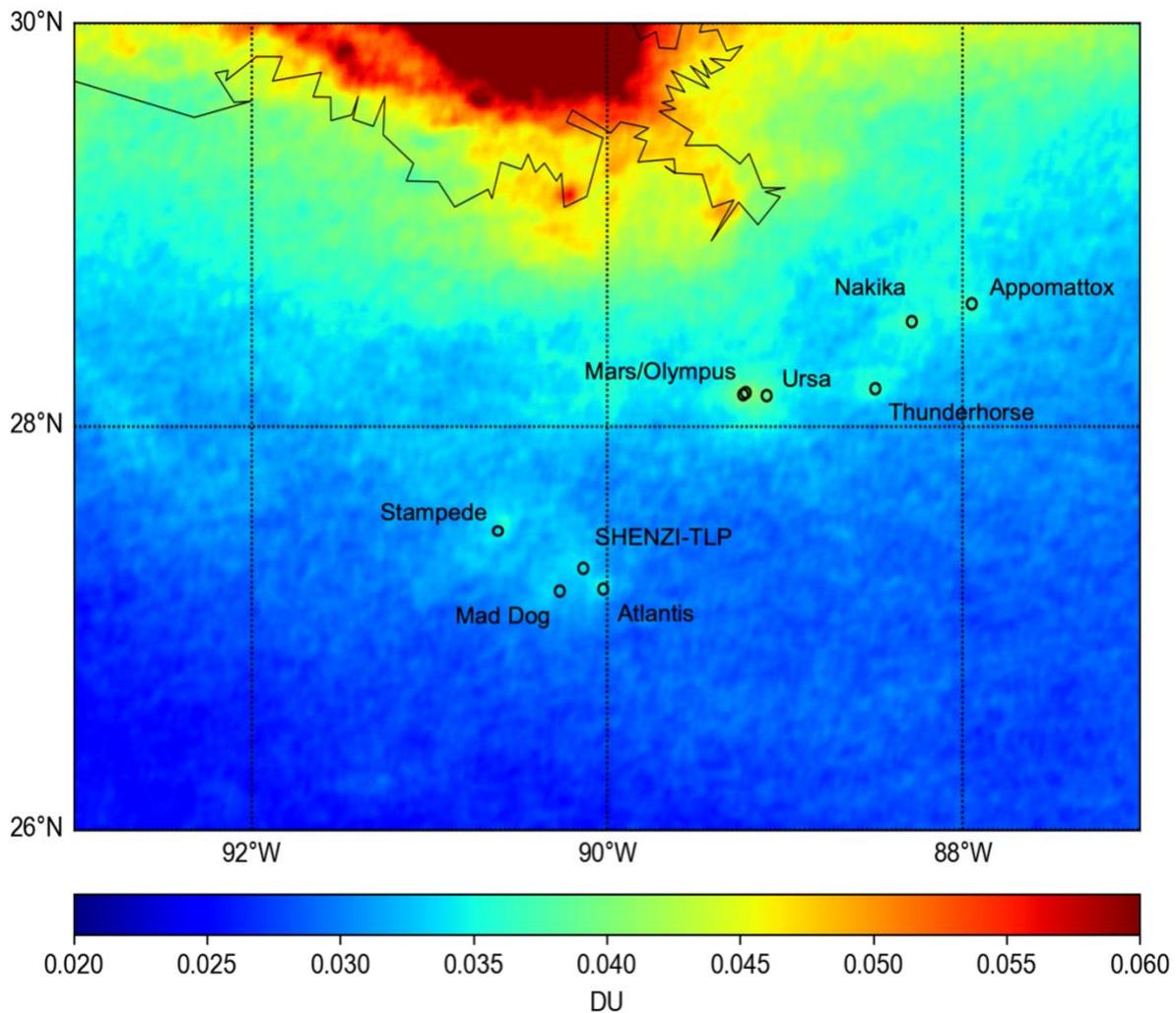
398



399 **Figure 7:** TROPOMI tropospheric NO₂ column averages for (a) 2019, (b) 2020, (c) 2021 and
 400
 401 (d) 2022. The averages were calculated using only the days on which MERRA-2 950 hPa winds
 402 at 18 UTC were less than 5 ms⁻¹. Pixels with a QA value of less than 0.75 were excluded during
 403 the averaging process.

3.3 Hotspots identified by TROPOMI

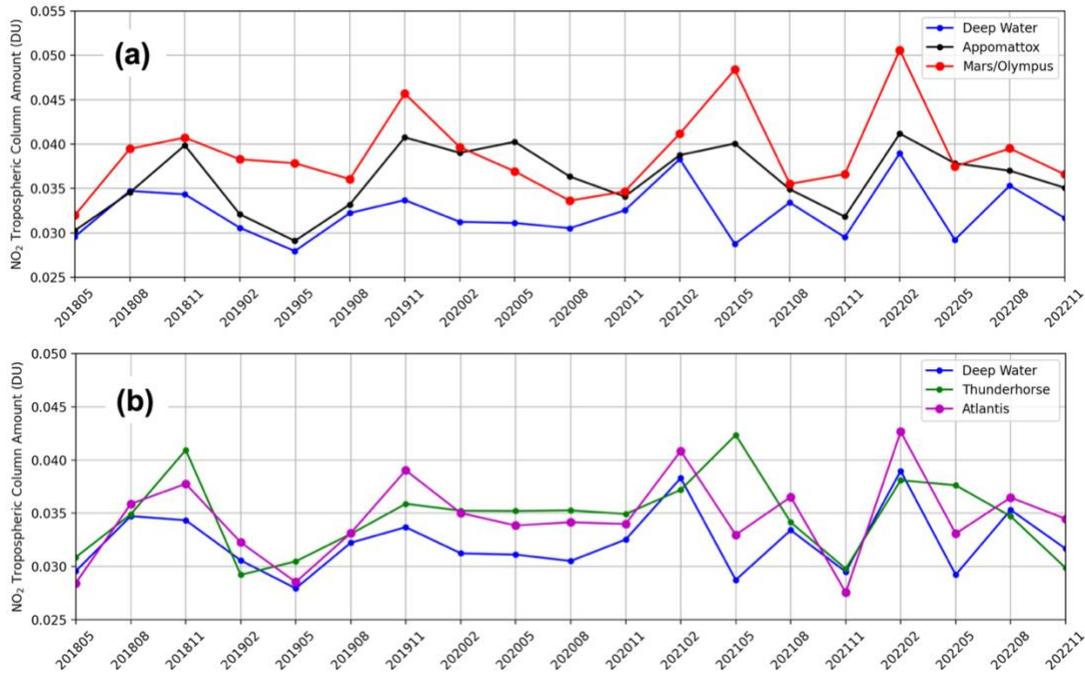
404
 405
 406 Aside from analyzing the long-term satellite record of the three areas, we also aimed to
 407 assess how well TROPOMI can observe ONG hotspots. The maps in Figure 7a-d show
 408 TROPOMI TrC NO₂ averages for each year in 2019-2022. These averages were calculated using
 409 only days for the calm wind case (MERRA-2 winds < 5 ms⁻¹) because this yields the best chance
 410 at isolating NO₂ hotspots from the surrounding areas. One noticeable difference from year to
 411 year is the varying levels of TrC NO₂, both over background and polluted regions. For instance,
 412 2021 had an overall higher background and near shore NO₂ amounts than other years. In 2020,
 413 the average offshore amounts were lower, possibly due to restrictions of the COVID-19
 414 pandemic (Bauwens et al., 2021; Fioletov et al., 2022). We identified several hotspots from these
 415 maps, shown as an average from mid-2018 through 2022 (Figure 8). The NO₂ hotspots
 416 correspond to one or more platforms in the 2017 BOEM emissions inventory.



418
 419 **Figure 8:** 2018-2022 tropospheric NO₂ column averages from TROPOMI, using only days on
 420 which MERRA-2 950 hPa winds at 18 UTC were less than 5 ms⁻¹. The circles and labels show
 421 locations of key platforms contributing to the NO₂ column hotspots.
 422

423 The largest hotspot consists of the Mars and Olympus platforms, both located around
 424 89.22° W and 28.17° N. It is visible clearly in every map in Figure 7. Ursa [89.104° W, 28.154°
 425 N], a platform roughly 10 km to the east, also contributes to this hotspot. Other platforms which
 426 can be identified are Thunderhorse [88.496° W, 28.19° N], Nakika [88.289° W, 28.521° N], and
 427 Appomattox [87.95° W, 28.61° N]. Appomattox only began operations in May 2019, and is
 428 visible for every year since 2019. Although located further south in the deepwater region, the
 429 Atlantis platform [90.027° W, 27.195° N], in addition to Mad Dog [90.269° W, 27.188° N] and
 430 SHENZI-TLP [90.135° W, 27.301° N] platforms all form visible hotspots in the same region on
 431 the TROPOMI maps. The hotspot to the northwest of Mad Dog, seen clearly at 90.6° W and
 432 27.5° N is from Stampede, another deepwater platform. It began production in 2018 and so it,
 433 like Appomattox, is not included in the BOEM 2017 inventory. The aforementioned platforms
 434 are all in the top twenty largest NO_x emitters in the GOM according to the BOEM 2017
 435 inventory, with exception of Stampede and Appomattox, the newer platforms that are not in the
 436 inventory. Numerous shallow water platforms are located above 28.5° N closer to the coast;

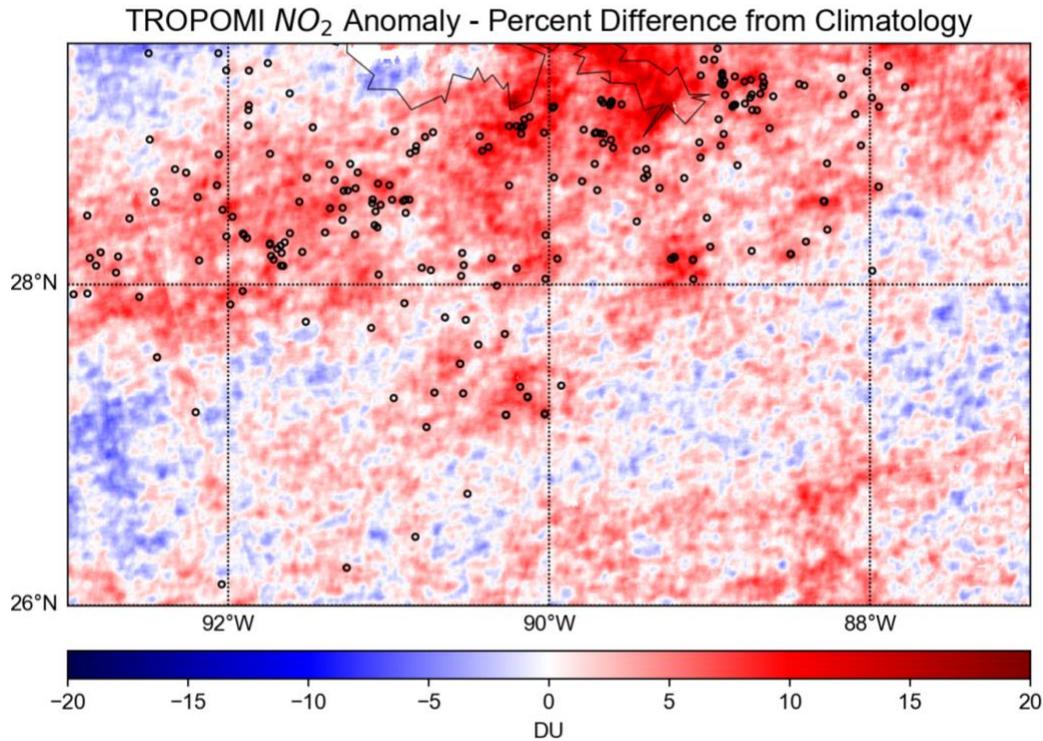
437 however, the individual platforms are generally low-NO_x emitters and are more difficult to
 438 distinguish from the background NO₂ values in the near-shore area. The Mars/Olympus hotspot
 439 represents a 25% increase above background levels of the deepwater area. Other platforms had
 440 smaller enhancements: Appomattox, Stampede, Thunderhorse and Atlantis had 13%, 7.7%, 8.1%
 441 and 6.9% higher NO₂ than background levels respectively. The hotspot enhancements were
 442 calculated by comparing the wind-based time series of each specific hot spot with the deepwater
 443 area time series (Figure 9).



444 **Figure 9:** Three-month time series for (a) the Appomattox and Mars/Olympus hotspots
 445 compared with the deepwater box (green box in Figure 2), and (b) Atlantis and Thunderhorse
 446 hotspots compared with the deepwater box. All of the time series were calculated with the calm
 447 wind case evaluated at each respective hotspot.
 448

449
 450 Time series for the hotspots were calculated with the same method as before, except the
 451 TROPOMI pixels used were restricted to within ± 0.1 degrees longitude and latitude of the
 452 platform coordinates. Given the distance of deepwater platforms from shore and the NO₂ column
 453 amounts being relatively small compared to that of polluted land areas, it is doubtful they
 454 produce any significant effects on coastal air quality. For reference, the Mars/Olympus hotspot
 455 maximum three-month value is only around 0.05 DU (Figure 9a). Nonetheless, observing these
 456 hotspots is important for evaluating the NO₂ budget over the GOM and ultimately validating the
 457 NO_x emissions inventories in the future. The NO₂ calm wind anomalies (Figure 10) provide
 458 another way to visualize the hotspots. Figure 10 is the average TrC NO₂ percent difference for
 459 each 0.01° by 0.01° grid cell between each calm wind day and its respective NO₂ monthly
 460 climatology. The same hotspots are visible in the map, the largest of which is Mars/Olympus.
 461 The percent anomaly for this hotspot is 9.8%, meaning that calm winds cause the accumulation
 462 of an additional 9.8% TrC NO₂ over Mars/Olympus compared to other days. The second and
 463 third largest calm wind anomalies are Nakika and Atlantis with 8.5% and 7.8% respectively. A
 464 large band of positive NO₂ anomalies stretches across the shallow waters over the area with a

465 high density of platforms (Figure 10). This is easier to see in the anomaly map rather the overall
466 TROPOMI NO₂ average in Figure 9.



467
468 **Figure 10:** Average TrC NO₂ anomalies from TROPOMI corresponding to calm wind days. The
469 top 250 NO_x emitting platforms from the BOEM 2017 inventory are plotted on the map with
470 empty circles. The TROPOMI data record from May 2018 through December 2022 was used in
471 the calculation and the anomalies were calculated with respect to monthly climatology.
472

473 474 475 **4. Conclusions**

476
477 We examined the 18+ year record of OMI satellite TrC NO₂ in the GOM region. Three
478 areas were considered for time series analysis: 1) Houston urban, 2) near shore and 3) deep
479 water. A trend analysis on the time series revealed a negative NO₂ trend for the Houston and near
480 shore areas and a slight increasing trend for the deepwater area. The average column amount of
481 the time series for Houston (0.148 DU) was three times greater than that of the near shore area
482 (0.0434 DU) which indicates the air over water is clean in comparison, despite the presence of
483 offshore ONG activity. The wind-classified time series showed that ONG activity does have an
484 impact on the NO₂ amount in deepwater region. For instance, in the calm wind case the NO₂
485 columns were around 33% higher on average than the onshore (wind from the south) case. The
486 calm wind trend for the deepwater area was +0.0021 DU per decade, indicating that there could
487 be a slight increase in NO_x emissions from deepwater ONG platforms since 2005.

488 We also showed the capability of TROPOMI to observe NO₂ hotspots from oil and
489 natural gas sources in the GOM. On average TROPOMI calm wind case maps, there are clear
490 indications of several NO₂ hotspots in the vicinity of ONG platforms. Visually this is observed
491 mostly in deepwater regions where the background is low enough for the hotspots to be isolated

492 from the background. The largest hotspot, Mars/Olympus, is 25% above the deepwater
493 background value partly because the two platforms are located only 1.9 km apart. The NO₂
494 anomalies from monthly climatology during calm wind conditions help quantify relative NO₂
495 enhancements from ONG operations. Clearly visible are distinct hotspots. Mars/Olympus is the
496 highest, with a 9.8% anomaly for the 2018-2022 TROPOMI record. Positive anomalies are also
497 observed in the shallow water area (north of 28° N) where there are numerous smaller platforms.
498 These platforms, while not clearly identifiable in the TROPOMI NO₂ averages, emit enough
499 NO_x to cause increases above the background values during calm winds.

500 Given that NO₂ enhancements from emissions can be seen by TROPOMI, a major
501 component of future work will focus on estimating emissions from these hotspots to validate
502 BOEM's ONG NO_x emissions inventories. Liu et al. (2022) and Goldberg et al. (2022) have
503 shown that this can be done without chemical models, using appropriate meteorological data and
504 when a sufficient source signal exists that can be identified in the satellite observations.
505 Presumably their approach can be applied to upcoming data from the TEMPO instrument, which
506 is the first geostationary UV-Vis instrument measuring NO₂ over North America. For validation
507 of TEMPO we will conduct a SCOAPE-II cruise in 2024 with Pandora and in-situ measurements
508 as in Thompson et al. (2023). The work presented here provides the first insight into long-term
509 trends over the GOM and demonstrates the capability of higher-resolution satellite instruments to
510 observe NO₂ hotspots, even over sources with comparatively smaller emissions and spatial
511 footprint than urban areas. Note that TEMPO can monitor ONG emissions as well as mobile
512 marine and land-based NO_x sources hour-by-hour throughout the GOM region. These processes
513 interact throughout the boundary layer NO₂ (Sullivan et al., 2023), contributing to the cycling of
514 reactive nitrogen across a range of environments, e.g., urban business, residential, shipping lanes,
515 ports, the vast petrochemical enterprise and wetlands.

516

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526

527 **Data Availability**

528 OMI total and tropospheric column NO₂ data can be downloaded from NASA GES DISC at
529 https://aura.gesdisc.eosdis.nasa.gov/data/Aura_OMI_Level2/OMNO2.003
530 (<https://doi.org/10.5067/Aura/OMI/DATA2017>; Krotkov et al., 2019). The high resolution OMI
531 dataset is a research data product developed by Lok Lamsal and can be downloaded from:
532 https://avdc.gsfc.nasa.gov/pub/data/satellite/Aura/OMI/V03/L3/OMNO2d_HR/OMNO2d_HRM/
533 TROPOMI tropospheric NO₂ column data is obtained from the Copernicus Hub at
534 <https://scihub.copernicus.eu/> (<https://doi.org/10.5270/S5P-s4ljg54>; Copernicus Sentinel-
535 5P, 2019 & 2021). The MERRA-2 data are available from GES DISC at
536 https://disc.gsfc.nasa.gov/datasets/M2I3NPASM_5.12.4/summary
537 ([doi:10.5067/QBZ6MG944HW0](https://doi.org/10.5067/QBZ6MG944HW0); GMAO, 2015). All analyses and creation of figures were
538 performed using publicly available Python modules.

539

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