

Satellites reveal a 28% drop in Ukraine’s Nitrogen oxides emissions during the Russia-Ukraine war in 2022
Inversion-based assessment of anthropogenic NO_x emission changes in Ukraine during the 2022–2023 war using TROPOMI satellite data

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15 **Abstract.** The outbreak of the Russia–Ukraine war in 2022 brought a huge impact on the Ukrainian ~~societ~~economic
~~production~~. We used TROPOMI NO₂ observations to constrain the EDGAR inventory and inverted daily anthropogenic NO_x
emissions in Ukraine from 2019 to 2023. Our results reveal a 15% reduction in Nitrogen oxides (NO_x) emissions during the
2022 war and an 8% reduction in 2023, both substantially exceeding the decrease caused by the COVID-19 pandemic.
Emission anomalies closely tracked the timing and location of major military actions, highlighting the sensitivity of NO_x
20 emissions to conflict-related disruptions. Regionally, Eastern Ukraine experienced larger reductions in NO_x emissions in both
2022 and 2023 by 29% and 17%, respectively, due to direct damage from frontline military operations. In contrast, Western
Ukraine experienced a relatively modest emission reductions of only 8% in 2022 with emissions increasing in some regions.
In 2023, the emissions increased in most western regions. After the outbreak of the war, the industrial sector experienced the
largest reductions in NO_x emissions, with decreases of 34% and 24% in 2022 and 2023, respectively, followed by the residential
25 sector, which fell by 23% and 18% respectively. War activities also emitted large amounts of NO_x, and such emissions partially
offset the emission reduction caused by the impact of war on socio-economic. By filtering out high-frequency emission
fluctuations induced by wartime activities through LOESS smoothing, our inversion results suggest that war-related emissions
may have offset approximately 8% and 10% of the anthropogenic NO_x reductions in 2022 and 2023, respectively. After
removing the war-related emissions, the inverted emission declines exhibit strong agreement with bottom-up emission
30 inventories and reported economic performance metrics. These findings provide new insights into the environmental and socio-
economic impacts of armed conflict.

To quantify this effect, we invert the anthropogenic Nitrogen oxides (NO_x) emissions in Ukraine from 2019 to 2022, a key indicator of human activities, to reflect the disruption of activities in different economic sectors due to war. We found a 28% decline in NO_x emissions during the war, if compared with the base year, which significantly exceeded the decrease caused by the 2020 COVID-19 pandemic. Eastern Ukraine experienced a 34% decrease in NO_x emissions, whereas the other regions experienced a decrease of 24%. The destruction of infrastructure and energy shortages severely impact the sustainable development of such social activities as industry, housing and transportation in Ukraine. These findings highlight the severe disruption of socio-economic activities due to the war, offering crucial insights into the broader implications of war on environmental and economic stability.

1 Introduction

In recent years, localized conflicts have proliferated, posing persistent challenges to economic stability, infrastructure, and social sustainability (Esteban et al., 2012; Gutierrez et al., 2024; Hou et al., 2024). Accurately and swiftly assessing the immediate and long-term impacts of modern warfare on human society has become increasingly critical. The ongoing Russia–Ukraine war, one of Europe's most significant conflicts since World War II (Adekoya et al., 2022), provides a unique lens for studying the consequences of contemporary warfare.

Since its outbreak on February 24, 2022, this conflict has profoundly disrupted Ukraine's economy, environment, and societal well-being (Ialongo et al., 2023; Kussul et al., 2023; Malarvizhi et al., 2023; Shumilova et al., 2023). Preliminary assessments of the damage caused by the conflict have already been made by a number of organizations. As of early 2023, the Office of the United Nations High Commissioner for Human Rights (OHCHR) reported nearly 22,000 civilian casualties (OHCHR, 2023). The war has significantly impacted all sectors, with the destruction of infrastructure, residential areas, and industrial facilities particularly severe. The resulting disruptions have not only affected Ukraine but also had global repercussions, including food shortages and energy export restrictions, which have in turn altered global energy and food systems (Carriquiry et al., 2022; Chen et al., 2023; Huang et al., 2023; Mottaleb et al., 2022; Rawtani et al., 2022). These events have also contributed to the expansion of global cropland and the loss of biodiversity (Chai et al., 2024). However, the accuracy and completeness of the available assessment data remain uncertain. This uncertainty stems from significant challenges in acquiring reliable and up-to-date statistics due to the deteriorating information environment during the conflict, making it difficult to assess the spatial and temporal dynamics of the war's impacts (Dando et al., 2023; Hou et al., 2024; Mueller et al., 2021). Consequently, there is an urgent need for remote, real-time quantitative methods to assess the extent of destruction across various regions at different stages of the conflict. Such approaches are crucial for supporting the well-being of civilians affected by the war.

Nitrogen oxides ($\text{NO}_x = \text{NO} + \text{NO}_2$) are significant air pollutants that reflect shifts in energy consumption, thereby serving as indicators of economic growth (Bilgen, 2014). Anthropogenic NO_x emissions primarily arise from fossil fuel combustion

activities (Miyazaki et al., 2012; Zhang et al., 2023b). Emissions sources include energy consumption in residential settings (e.g., natural gas) (Lebel et al., 2022), industrial production (Gholami et al., 2020; Li et al., 2023b; Zhu et al., 2023), energy supply (e.g., power plants) (Islam et al., 2023; Liu et al., 2015; Tang et al., 2019), and agriculture (Chen et al., 2018; Liu et al., 2018). With increasing motorization and urbanization, the transportation has emerged as the largest contributor to NO_x emissions (Wu et al., 2017), accounting for approximately 42% of total NO_x emissions in Europe (Sun et al., 2018). Consequently, NO_x emissions often represent changes in the intensity of activities in various economic sectors. As a short-lived gas, directly emitted nitric oxide (NO) rapidly oxidizes to form nitrogen dioxide (NO₂), which, in the presence of sunlight and oxidized volatile organic compounds (VOCs), contributes to net ozone (O₃) generation (Chameides, 1978; Crutzen, 1970). This short-lived nature typically results in a strong correlation between atmospheric NO₂ concentrations and NO_x emissions in localized regions (Richter et al., 2005). This relationship provides a theoretical basis for the grid-scale inversion of NO_x emissions from satellite-observed atmospheric NO₂ data (Reuter et al., 2014).

The characteristics of satellite data, including their large-scale coverage and timeliness, ensure effective data support even under unforeseen circumstances, providing valuable insights into the spatial pattern and scale of global NO_x emissions through top-down inversion (Li et al., 2023a). Various inversion techniques, including plume, Gaussian, and box models, permit the estimation of surface NO_x emissions directly from satellite-derived NO₂ vertical column density (VCD) (Beirle et al., 2011; Benjamin de Foy et al., 2014; Duncan et al., 2013). Moreover, approaches that integrate satellite observations with atmospheric chemistry transport models, such as mass balance methods, 4D-Var methods, and ensemble Kalman filtering methods, facilitate a more accurate characterization of spatial and seasonal emission trends (Gu et al., 2016; Martin et al., 2003; Miyazaki et al., 2012; Qu et al., 2017; Stavrakou et al., 2013; Xu et al., 2013). Studies have revealed the impact of the COVID-19 pandemic on human activity and the spatial and temporal dynamics of regional economies based on inverted NO_x emissions from satellite observations (Feng et al., 2020; Mao et al., 2024; Miyazaki et al., 2021; Zheng et al., 2021; Guevara et al., 2021; Li et al., 2023a; Li and Zheng, 2023; Luo et al., 2023). Previous studies have demonstrated that satellite-based NO₂ observations can serve as effective indicators of the societal impacts induced by regional conflicts. For instance, Lelieveld et al. (Lelieveld et al., 2015) analyzed a decade of NO₂ data from the OMI satellite and reported that widespread displacement caused by armed conflicts in the Middle East resulted in significant deviations in anthropogenic NO_x emissions from their typical interannual trends. More recently, Zhang et al. (Zhang et al., 2023a) further observed pronounced fluctuations in NO₂ concentrations over Ukraine between February and July 2022, driven by changes in anthropogenic emissions, as revealed by TROPOMI satellite measurements combined with a machine learning approach. In light of the limitations of existing bottom-up inventories and the challenges of obtaining timely and reliable ground-based data during conflicts, there remains a critical need for robust, satellite-driven top-down approaches to accurately quantify changes in NO_x emissions in Ukraine during the war period as an indicator of the impact of the war on society. Analyzing changes in NO_x emissions in Ukraine during the war period facilitates a quantitative assessment of disruptions in industrial output, transportation activities, residential energy consumption, and broader implications for sustainable development.

In this study, we optimise the inversion framework we previously published (Mao et al., 2024) to invert changes in anthropogenic NO_x emissions during the ~~2022~~ war from 2022 to 2023 in Ukraine using satellite observations of NO₂ VCD from the TROPOspheric Monitoring Instrument (TROPOMI) satellite (Van Geffen et al., 2022). The framework utilizes the anthropogenic NO_x emission inventory from Emissions Database for Global Atmospheric Research (EDGAR) Community Emissions Data System (CEDS) anthropogenic NO_x emission inventory (Crippa et al., 2024) ~~(Hoesly et al., 2018)~~ as the prior emission and simulates the atmospheric NO₂ transport process using the GEOS-Chem atmospheric chemistry model (GEOS-Chem 14.0.0, 2022). We employ the finite-difference mass-balance (FDMB) method (East et al., 2022) to establish a linear relationship between anthropogenic NO_x emissions and satellite-derived NO₂ ~~column concentration~~ VCD observations, enabling us to invert anthropogenic NO_x emissions during the war in Ukraine ~~(see Materials and methods)~~. Our analysis captures the spatial and temporal variability of NO_x emissions across the industrial, agricultural, transportation, and residential sectors in Ukraine from 2019 to ~~2022~~ 2023. This approach allows us to assess the changing intensity of the war's impact on various economic activities within Ukraine. For comparison, we also analyzed NO_x emissions in 2020, which were affected by the pandemic. In addition, we divided Ukraine into eastern and western regions based on the scope of the war and estimated the NO_x emissions from different sectors in each region. By conducting a comparative analysis of NO_x emission changes across regions and sectors during the war, we identified the driving factors behind these changes, revealing the profound economic disruptions caused by the war. This analysis underscores the importance of sustainable development strategies and resilience planning in mitigating the impacts of such crises on modern society.

2. Materials and methods

2.1 Atmospheric Chemical Transport Model

We utilized version 14.0.0 of the GEOS-Chem model (Melissa, 2022), to conduct forward simulations of NO₂ VCDs in the troposphere over Ukraine, which has been widely used in the inversions of surface CO₂ flux (Wang et al., 2022), CH₄ (Shen et al., 2023) and NO_x emissions (Zheng et al., 2020). Specifically, local simulations for the European region were conducted within the geographical bounds of ~~30~~ 30°W–~~50~~ 45°E longitude and ~~30~~ 35°N–~~70~~ 65°N latitude. The boundary conditions were derived from global simulations conducted using the same version of the GEOS-Chem model. The model was operated at a horizontal resolution of ~~0.525~~ 0.525° × ~~0.625~~ 0.625° and a vertical resolution of 47 layers. Meteorological data from the GEOS forward processing Modern-Era Retrospective Analysis for Research and Applications, version 2 (MERRA-2GEOS-FP) was employed to drive the model, ~~while, the natural source NO_x emissions were based on the model's default emission inventory.~~

2.2 Prior Emissions

~~The anthropogenic NO_x emissions inventory used in this study originated from the Community Emissions Data System (CEDS) developed by the Pacific Northwest National Laboratory and the University of Maryland's Joint Institute for Global Change Research (Hoesly et al., 2018). In the inversion process, the inventory was repurposed by reallocating emissions across sectors.~~

~~Specifically, emissions from industrial sources and sporadic energy consumption were combined into a single category. Emissions from the residential sector were derived from the residential, commercial, and waste disposal sources outlined in the prior inventory. The agricultural and transport sectors remained consistent with the initial inventory. It is also pertinent to~~
130 ~~note that emissions from ships and aviation were excluded from the prior inventory, which aligns with this study's focus on anthropogenic emissions originating from land sources. The anthropogenic NO_x emissions inventory utilized in this study is derived from the Emissions Database for Global Atmospheric Research (EDGAR v8.1), developed by the European Commission's Joint Research Centre (JRC) (Crippa et al., 2024). This inventory estimates national total emissions by sector based on international energy and activity statistics combined with emission factors, following the same methodology as earlier~~
135 ~~versions of EDGAR (Crippa et al., 2018). Emissions are spatially distributed at a resolution of 0.1° × 0.1° using dynamic spatial proxies. For point sources such as power plants and industrial facilities, location information is obtained from the Global Energy Monitor database, and missing data are supplemented using non-residential building surface data from the Global Human Settlement Layer (GHSL). Line sources, such as international shipping, are derived from the Ship Traffic Emission Assessment Model (STEAM). Area sources, particularly residential combustion, are distributed based on population density (GHS-POP R2023A) weighted by heating degree days (HDD) to account for temperature-dependent fuel consumption.~~

In ~~addition, we also used emissions from natural sources, including biomass burning and soil.~~~~this study, the~~~~The~~ biomass burning NO_x emissions data was derived from the GFED4 inventory, which is based on the Global Fire Emissions Database (GFED)(Randerson et al., 2018). GFED4 provides global data on monthly burned area at a 0.25° spatial resolution, using a
145 combination of MODIS burned area maps, active fire data from the Tropical Rainfall Measuring Mission (TRMM), and the Along-Track Scanning Radiometer (ATSR) sensor family(Giglio et al., 2013). This inventory offers significant improvements over previous versions by incorporating a higher spatial resolution and more accurate fire mapping, which is critical for estimating emissions from biomass burning. The GFED4 inventory accounts for the dynamics of biomass burning and the associated emissions, including NO_x, by mapping burned areas at fine spatial and temporal scales. This allows for the
150 assessment of interannual variability and long-term trends in biomass burning. The inclusion of biomass burning emissions is essential for understanding the contributions of wildfires and other biomass combustion sources to global NO_x levels, especially in regions affected by seasonal fires.

The soil NO_x emissions used in this study are based on a modified version of the Berkeley-Dalhousie Soil Nitrogen Oxide Parameterization (BDSNP), originally developed by Hudman et al. (2012) and implemented in GEOS-Chem. This emission
155 inventory represents a significant advancement over previous parameterizations by adopting a more mechanistic approach to modeling soil NO_x emissions. The BDSNP accounts for the persistent dependence of emissions on soil moisture and temperature, as well as the pulsed emissions following soil wetting events, which are critical for capturing the temporal variability of soil NO_x emissions. The inventory also includes detailed spatiotemporal representations of nitrogen inputs from fertilizers, manure, and atmospheric deposition.

160 2.3 Satellite NO₂ VCD Observations

Currently, the TROPOMI satellite provides high-quality NO₂ VCD data with enhanced spatial resolution and signal-to-noise ratio, which are beneficial for NO_x emission studies (Sekiya et al., 2022; Veeffkind et al., 2012). In this study, we employed the most recent versions of the TROPOMI NO₂ product to provide optimal observational constraints for the inversion framework. Specifically, we used the v2.4.0 official reprocessed dataset for the period from 1 January 2019 to 25 July 2022,
165 the v2.4.0 official offline dataset from 26 July 2022 to 12 March 2023, the v2.5.0 official offline dataset from 13 March 2023 to 26 November 2023, and the v2.6.0 official offline dataset for the remaining days of 2023. These products incorporate improved Level-1b processing and retrieval algorithms and represent the most up-to-date and consistent TROPOMI NO₂ datasets available. All data were obtained from the TEMIS portal (TEMIS, 2025Tropospheric Emission Monitoring Internet Service).~~In this study, we employed TROPOMI v2.3.1 data to provide observational constraints for our inversion framework.~~
170 TROPOMI, launched in October 2017 onboard the European Space Agency (ESA) S5P spacecraft, provides high-quality global daily observations that exhibit a strong correlation with ground-based data (Ialongo et al., 2020) and demonstrate relative stability in statistical uncertainty (Van Geffen et al., 2020). TROPOMI is a UV–visible spectrometer aboard the Sentinel-5P satellite in a sun-synchronous polar orbit, crossing the equator at approximately 13:30 local time. -We screened GEOS-Chem simulations that overlapped with the transit time of the TROPOMI grid to participate in the inversion.~~We meticulously~~
175 ~~screened grids with daily observations of NO₂ VCDs in the troposphere using TROPOMI.~~ This process ensured that the data quality threshold exceeded 0.75-5, ~~and the cloud cover was below 30%.~~ TROPOMI data were gridded using the HARP toolkit of the Coordinated Toolkit for Scientific Earth Observation Data (CTSOD), which facilitated the amalgamation of daily global observations and their alignment to the same spatial resolution as the simulated concentrations. We synchronized the concentration values for each grid of the simulated concentrations with the moment of TROPOMI transit to ensure spatial and
180 temporal coherence between the two datasets.

In this study, we quantified relative changes in TROPOMI NO₂ VCDs between the wartime period in 2022 and full year of 2023 and the corresponding periods in pre-war baseline (mean of 2019 and 2021; excluding 2020 due to COVID-19 anomalies, same thereafter) (Fig. 1). Results demonstrate that satellite NO₂ VCDs effectively capture spatiotemporal variability of air pollution during the war, though their representation of ground-level emissions remains limited given their tropospheric
185 column nature.

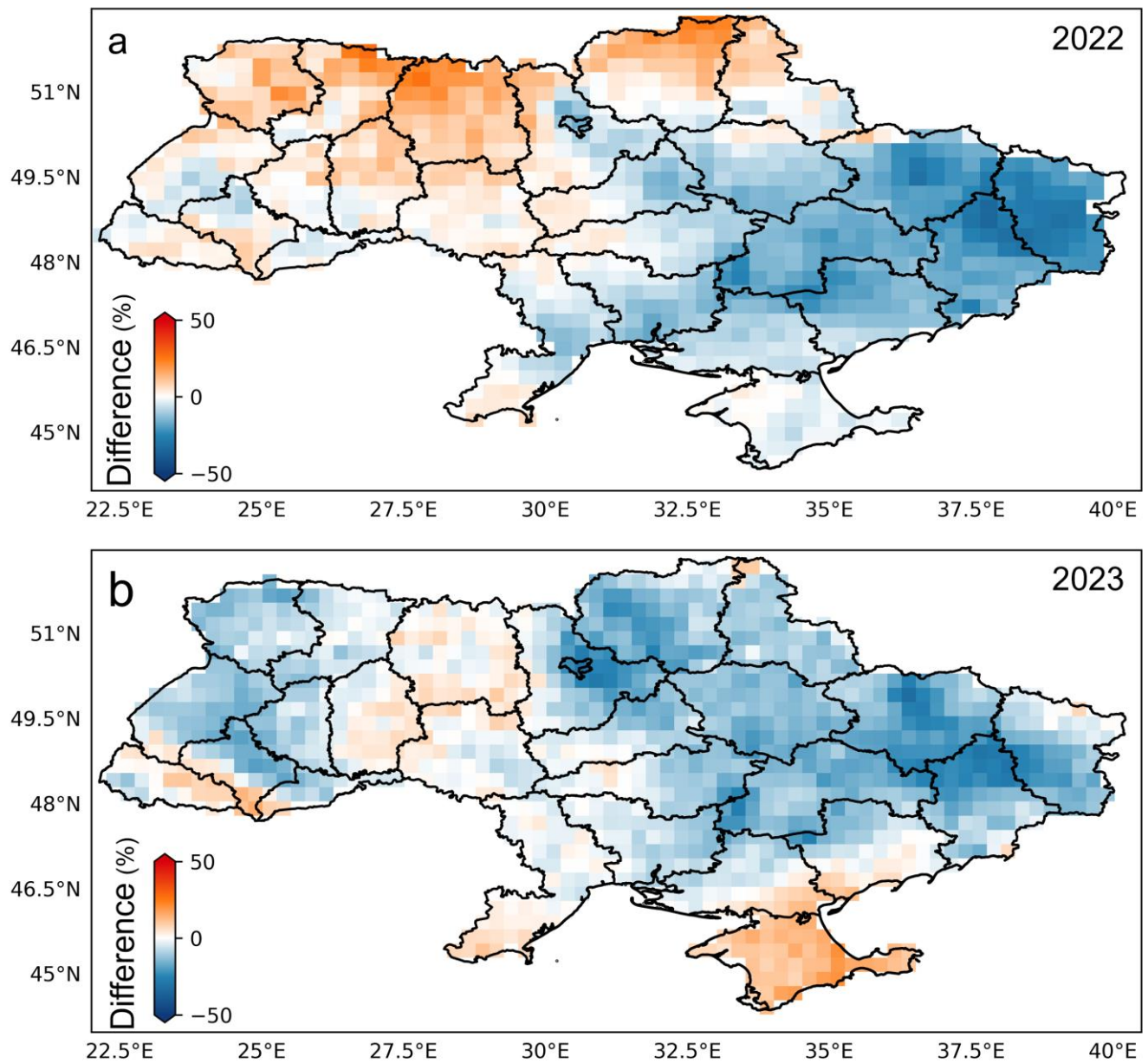


Fig. 1 The Spatial changes in satellite-observed NO₂ VCDs are illustrated (a) during the war (March to December) in 2022 and (b) in 2023 relative to the corresponding periods of baseline (average of 2019 and 2021).

Furthermore, when using TROPOMI NO₂ VCD as observational constraints for inversion, the quality of satellite observations directly impacts the inversion accuracy. In Ukraine, the lack of sufficient valid pixels severely hampers the reliability of daily inversions. We applied a multi-day moving average window to constrain the daily simulated concentrations to address this

limitation. The effectiveness of this approach has been demonstrated in previous studies (e.g., Zheng et al., 2020). We evaluated the number of valid TROPOMI observation pixels over Ukraine using different moving average window lengths during the study period (Fig. S1a). When the window size reached 7 days, the data loss significantly decreased, and the proportion of valid grid cells exceeded 90%. However, applying a multi-day average may suppress short-term fluctuations in NO₂ concentrations, potentially limiting the ability to capture rapid changes associated with wartime dynamics. We compared the NO₂ VCDs over Ukraine using 7-day and 11-day moving averages (Fig. S1b). The results showed no substantial difference between the two, but the 11-day average tended to smooth out peak concentrations more strongly. Based on this assessment, we adopted the 7-day moving average of TROPOMI NO₂ VCDs to constrain the model simulations in this study. ~~We summarized the number of valid grids and their proportion relative to the total grid count for daily TROPOMI observations over Ukraine (Fig. S1). The results show a significant reduction in valid observations during winter, with the proportion falling below 40%. To address this, we averaged the satellite data to the monthly scale, which improved the valid grid proportion to over 90% throughout the study period (Fig. S1). Therefore, we conducted the inversion of anthropogenic NO_x emissions on a monthly scale to minimize the impact of missing satellite observations.~~

2.4 Inversion method

In the previous study, we developed a two-step inversion framework for estimating global anthropogenic NO_x emissions (Mao et al., 2024). In this study, we optimized this framework by incorporating the treatment of ~~natural source~~-NO_x emissions from natural sources and used ~~them~~ it to invert ~~monthly~~ daily anthropogenic NO_x emissions in Ukraine from 2019 to ~~2022~~ 2023. This inversion framework is based on the FDMB method and draws on the framework developed by Zheng et al. (2020), which optimizes the prior anthropogenic NO_x emissions by fitting the simulated NO₂ concentrations to satellite NO₂ VCD observations and interannual variability ~~to simulated NO₂ concentrations~~.

We output hourly NO₂ concentrations from the GEOS-Chem model and sampled the values corresponding to the satellite overpass times. Using pressure of each model layer, we vertically regridded the model output to match the vertical resolution of the TROPOMI satellite. Following the recommendations in the TROPOMI User Manual (van Geffen et al., 2019), based on the vertical level of the tropopause provided in the satellite product and the tropospheric averaging kernel, we integrated the tropospheric NO₂ profiles to calculate the simulated NO₂ VCD for each model grid cell.

~~The FDMB method uses a prior NO_x emission inventory and simulates two scenarios using an atmospheric chemical transport model: a baseline simulation and a reduced emission simulation, where emissions are scaled by a certain factor. By comparing the NO₂ VCD differences between these two simulations, we derive a scaling factor that relates NO_x emissions to NO₂ VCDs, establishing a local correspondence between the two. This allows the conversion of the NO₂ VCD bias between satellite observations and prior simulations into an error estimate for the prior emissions. The FDMB method optimizes NO_x emissions by coupling simulated atmospheric NO₂ VCD driven by prior emission inventories, with satellite observations through a specific scaling factor. This scaling factor is derived by comparing simulated NO₂ VCDs under identical meteorological~~

conditions using prior and reduced-emission scenarios within the CTM framework. It quantifies the sensitivity of NO₂ VCDs to NO_x emission perturbations and establishes a localized relationship between the two. This enables the transformation of discrepancies between satellite-retrieved and model-simulated NO₂ VCDs into estimates of errors in the prior emissions.

In this study, for the ~~2019~~ simulation in 2019~2022, we used the ~~2019~~ "bottom-up" CEDS-EDGAR NO_x emission inventory as the prior inventory and simulated both the prior baseline (1×) and 40% emission reduced NO₂ VCDs using the GEOS-Chem model ~~CTM~~. From this, we estimated the daily emission scaling factor β_{12019} ~~for each month of 2019~~ (Eq. 1):

$$\beta_{12019} = 0.4 \div \frac{\Delta\Omega}{\Omega_{base\ sens}} \quad (1)$$

Where $\Delta\Omega = \Omega_{base} - \Omega_{sens}$, with Ω_{base} being the baseline-prior simulated VCD concentration and Ω_{sens} the scaled simulation of VCD concentration. Using the scaling factor β_{12019} , we constrained the ~~2019~~ prior emission in 2019~2022 ($E_{prior-2019}$) using TROPOMI observations ~~to simulate NO₂ concentrations and inverted~~ to obtain the posterior NO_x emissions ~~for 2019, ($E_{post-2019}$)~~ (Eq. 2):

$$E_{post-2019} = (1 + \beta_{12019} \Delta\Omega_{bias-2019}) E_{prior-2019} \quad (2)$$

Where $\Delta\Omega_{bias-2019} = \frac{\Omega_{sate-2019} - \Omega_{base}}{\Omega_{base}}$ represents the relative bias between the prior VCD simulation ~~simulated concentrations~~ and satellite VCD observations. Due to the reduced TROPOMI observation coverage on specific days, as discussed in Sect. 2.3, we employed 7-day moving averages of satellite NO₂ observations for comparison with daily simulated VCDs in this study. Specifically, for each day during the study period, the simulated NO₂ VCDs were constrained using the quality-filtered TROPOMI NO₂ VCD observations. This approach has been demonstrated to be effective by Zheng et al (2020).

For the inversion of 2023, which is not covered by the prior inventory, we applied the previously developed two-step inversion framework. Using the FDMB method, we optimized the 2019 anthropogenic NO_x emissions inventory. However, since CEDS only updates the inventory up to 2019, the interannual variability in anthropogenic NO_x emissions is larger than the uncertainty in the prior inventory. Therefore, for emissions beyond 2019, we adopted a two-step inversion method to extend the emission inventory.

We used the 2022 prior emission inventory, combined with 2023 meteorological fields and natural source emissions, to simulate the prior NO₂ VCDs for 2023 ($\Omega_{simu,2023}$) using the GEOS-Chem model. This setup implies that the interannual variation in simulated NO₂ concentrations is primarily driven by non-anthropogenic factors. First, we used the 2019 CEDS emission inventory in the GEOS-Chem CTM to simulate NO₂ concentrations from 2020 to 2022 ($\Omega_{simu,202x}$), where the meteorological conditions and natural source emissions data were synchronized with the simulation period. This indicates that the changes in NO₂ concentrations after 2019 are primarily driven by non-anthropogenic factors (Zheng et al., 2020). Since the variations in satellite-observed NO₂ VCD represent the total changes, we calculated the interannual changes in NO₂ VCD caused by anthropogenic NO_x emissions ($\Delta\Omega_{annu}$) (Eq. 3); and converted these changes into interannual variations in

255 anthropogenic NO_x emissions using scaling factor of 2022, β_{2022} ~~2019~~. This allowed us to extend the ~~2019–2022~~ prior inventory to 2022–2023 ($E_{\text{prior}_2023\text{x}}$) (Eq. 4).

$$\Delta\Omega_{\text{annu}} = \frac{\Omega_{\text{sate},2023\text{x}}}{\Omega_{\text{sate},202219}} - \frac{\Omega_{\text{simu},2023\text{x}}}{\Omega_{\text{simu},202219}} \quad (3)$$

$$E_{\text{prior}_2023\text{x}} = (1 + \beta_{202219}\Delta\Omega_{\text{annu}})E_{\text{prior}_202219} \quad (4)$$

260 Where $\Omega_{\text{sate},2023\text{x}}$ and $\Omega_{\text{sate},202219}$ are the satellite-observed NO₂ VCDs ~~for the corresponding months in 2020–2022~~2023 and ~~2019–2022~~. The inherent uncertainties in the prior inventory were not optimized and may propagate into the emission inventory for ~~2020–2022~~2023. As a result, we repeated the inversion process for ~~2019–2019–2022~~ and simulated two scenarios for the ~~2020–2022~~2023 NO₂ VCDs. Using the FDMB method, we ~~inverted to obtain~~ed the posterior daily anthropogenic NO_x emissions ~~for each of the years~~.

265 For sectoral emissions, we allocated grid-level totals based on the sectoral distribution ratios provided in the EDGAR inventory. For the year 2023, due to the absence of updated sector-specific data, we assumed that the sectoral proportions in 2022 remained unchanged and applied them to the 2023 total emissions.

270 In this study, a key advancement of our framework is the explicit decoupling of natural and anthropogenic sources. Unlike conventional approaches that fix natural emissions as static inputs, we simulate scenarios with a 40% curtailment of emissions with the same percentage of curtailment of natural source emissions and apply the same inversion constraints to both source categories during inversion, ensuring unbiased attribution of observed NO₂ changes. Subsequently, its prior inventory scaling was used to differentiate NO_x emissions from natural sources from anthropogenic sources, effectively reducing crosstalk between natural variability and anthropogenic signals.

275 ~~In addition, we evaluated the changes in emissions from different sectors. Given the absence of data characterizing the spatial and temporal changes in emissions in each sector from 2020 to 2022, we estimated the sectoral allocation. In accordance with the methodology proposed by Zheng et al. (2020), the initial allocation of optimized emissions was based on the proportion of total emissions attributed to each sector in the previous emission inventory. The dominant sector in each grid was assumed to remain constant over time. However, of note, there is a possibility of spatial shifts in sectoral emissions over time. Therefore, we identified grids in Ukraine where sectors dominated, computed the emission changes in these regions relative to the prior inventory, and derived a relative change factor to adjust the allocation to each sector. This method accounts for both the spatial distribution of emissions by sector and temporal changes and offers a valuable approach for estimating emissions from different sectors in the absence of comprehensive data.~~

2.5 Uncertainty

285 To quantify the prior uncertainty remaining in the inversion system, we conducted an observing system simulation experiment (OSSE) (Atlas, 1997). In the OSSE, we used the ~~CEDS-EDGAR~~ emissions in ~~2019–2022~~ as the prior, assuming that the true emissions in ~~2019–2022~~ were ~~1–20.5~~ times the prior, and the true emissions in ~~2022–2023~~ were ~~0–71.25~~ times the ~~prior~~true values in 2022. The assumed true emissions for 2022 were informed by the average uncertainty of NO_x emissions over the

European Union as reported by the EDGAR inventory, which is approximately 51.7% (Crippa et al., 2018). The assumed true emissions for 2023 were guided by the interannual variability of TROPOMI NO₂ VCDs, which showed that monthly NO₂ concentrations in 2023 differed from those in 2022 by more than 25% in certain months. The pseudo-true values ~~for 2019 and~~

290 ~~2022~~—were simulated using the GEOS-Chem model and processed into the same grid distribution as the TROPOMI observations described in Section 2.3, serving as pseudo-observations. These pseudo-observations were then used to constrain the prior emissions in the inversion framework, allowing us to quantify the reduction in prior bias achievable by the inversion system. For comparison, we also applied the same pseudo-observations to constrain the inversion framework from our previous study (Mao et al., 2024) to evaluate the impact of including natural source emissions on the accuracy of anthropogenic NO_x emissions. Another major source of uncertainty in our results is the uncertainty in the satellite observations. The specified random uncertainty for individual TROPOMI tropospheric NO₂ VCD measurements is between 25% and 50%, with a precision of 0.7×10^{15} molec cm⁻² (Malyska et al., 2024). And TROPOMI NO₂ observations provide precision estimates for each grid. In our results, we converted the precision range of NO₂ VCD observations for each grid into a corresponding precision range for the anthropogenic NO_x emission constraints using the conversion factor β . We then summed the remaining prior uncertainty 300 for each grid, as computed in the OSSE, to determine the uncertainty range in the inversion results.

2.6 Evaluation of the Inverted Anthropogenic NO_x Emissions

Validating the accuracy of the inverted anthropogenic NO_x emissions is challenging due to the limited availability of independent anthropogenic NO_x emission inventories and surface observations specifically for Ukraine. To assess the robustness and reliability of the inversion results, we used a multi-pronged approach. First, we performed simulations of both 305 prior and posterior tropospheric NO₂ VCDs using the GEOS-Chem model, which accounts for atmospheric chemistry and transport processes. These simulated concentrations were then compared to the TROPOMI satellite-observed NO₂ VCDs. This comparison allowed us to assess the temporal and spatial consistency of the inversion framework against high-resolution satellite observations.

Additionally, as the scope of our inversion extended over the European land area, we leveraged independent in-situ 310 observations of surface NO₂ concentrations from the European Environment Agency (EEA). These ground-based observations served as a critical reference to validate the simulated surface concentrations corresponding to the posterior NO_x emissions. Specifically, we fitted the simulated surface NO₂ concentrations to the time series of in-situ measurements for multiple stations across Europe, facilitating an evaluation of how well the prior and posterior simulations captured the observed NO₂ variability. To quantify the accuracy of these fits, we performed a linear regression analysis comparing the time-averaged surface 315 concentrations from the modeled grid cells with the in-situ measurements. We then evaluated the fitting accuracy by examining the R² values and root mean square errors (RMSE) between the simulated and observed concentrations.

Moreover, to assess the temporal stability and robustness of the inversion framework, we compared the interannual variations in the relative deviations between the posterior results and both satellite and ground-based observations. By examining how

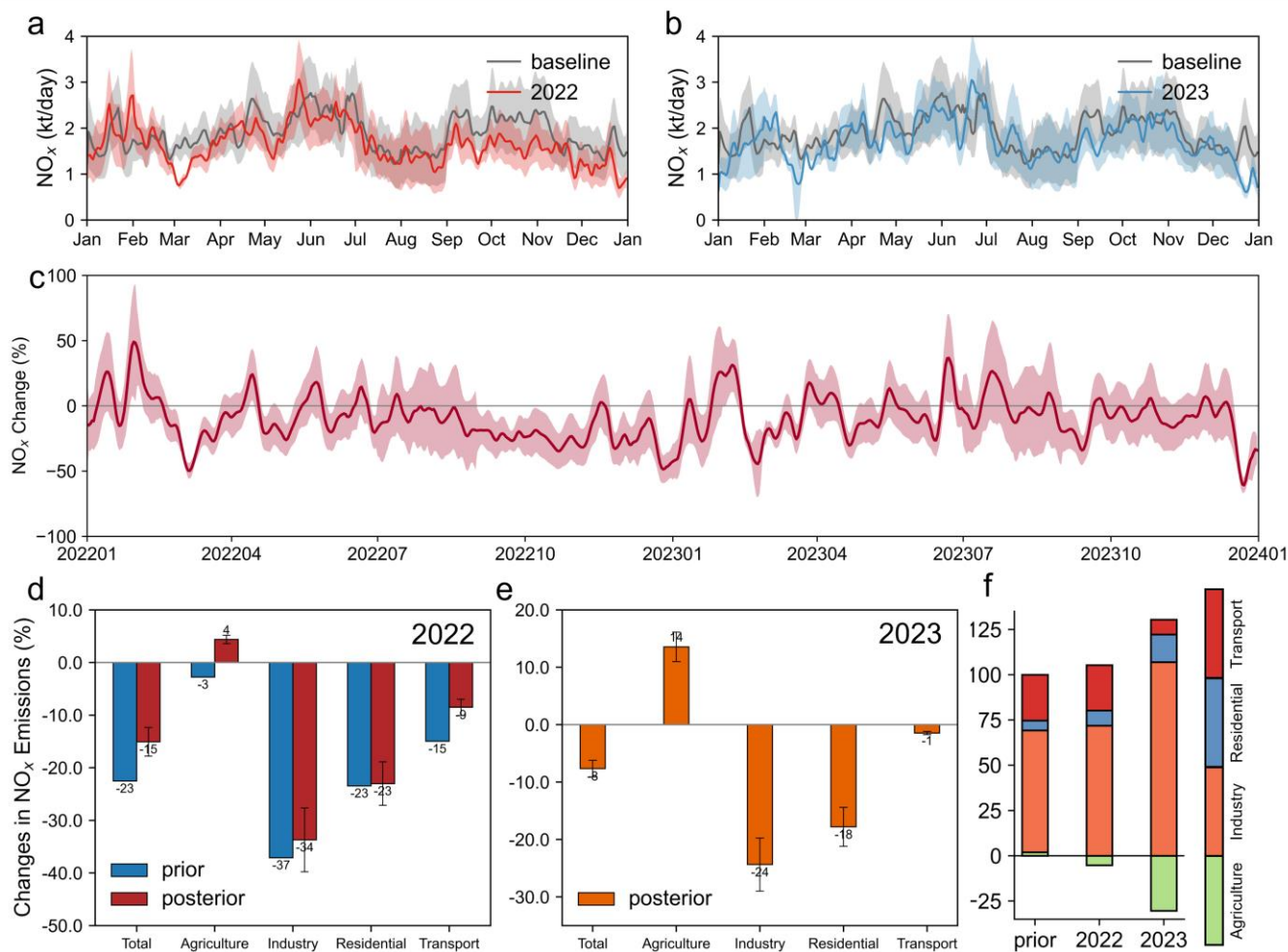
the relative biases evolved over time, we were able to ensure that the posterior emissions remained consistent with observational trends.

3. Results

3.1 National Decline

To ascertain the impact of the Russia–Ukraine war on anthropogenic NO_x emissions in Ukraine, we compared the ~~monthly~~ temporal variations in anthropogenic NO_x emissions during the war with the pre-war levels (baseline emissions). We derived baseline emissions by averaging the emissions in 2019 and 2021, excluding 2020 owing to the influence of the COVID-19 pandemic. Fig. 1-2 depicts the seasonal variations in NO_x emissions in 2022 and 2023 and their changes relative to the baseline ~~years~~. In 2022, Ukraine exhibited a 2412% ($\pm 7.62.1\%$) reduction in NO_x emissions compared with the baseline, and the reduction during the war period (starting on February 24~~March to December~~) was 2815% ($\pm 7.02.7\%$). The first sharp decline of 4125% ($\pm 164.6\%$) occurred in March, coinciding with the start of the full-scale Russian invasion in late February 2022, indicating a significant short-term societal disruption. Subsequently, the rate of decline in emissions slowed, stabilizing at approximately 157% ($\pm 3.81.2\%$) from June to ~~October~~August. This stabilization was due to the stalemate of the war in eastern regions of Ukraine, which suggests a rapid societal adaptation within a month of the war's outbreak, with partial restoration of social functions and preparations for long-term war. After ~~September~~October, the NO_x emissions showed a new round of decline of 23.8% ($\pm 4.3\%$), reaching a peak decline of 5530.5% ($\pm 245.5\%$) in December. This was primarily due to the increased energy demand in the baseline years and intensified energy shortages in 2022 during the cold season. Notably, although we cannot precisely quantify the contribution of population displacement to emission reductions, the continued outflow of residents due to the ongoing impacts of the war likely contributed to the enhanced decrease in wintertime NO_x emissions.

In 2023, anthropogenic NO_x emissions in Ukraine declined by 7.6% ($\pm 1.4\%$) relative to the baseline (Fig.2 b, c, e). The most pronounced reductions were observed in February, April, September, and December. Notably, unlike the sustained emission decline throughout 2022, 2023 exhibited intermittent increases, with higher emissions than the baseline observed in March and from May to July. From a seasonal perspective, persistent energy shortages during the cold season remained a critical constraint in early spring (before March) and late autumn (after September), contributing to continued emission reductions during these periods.~~The war induced constraints on energy supply and population migration caused the seasonal emission pattern in 2022 to differ significantly from that of the baseline (Fig. 1a).~~



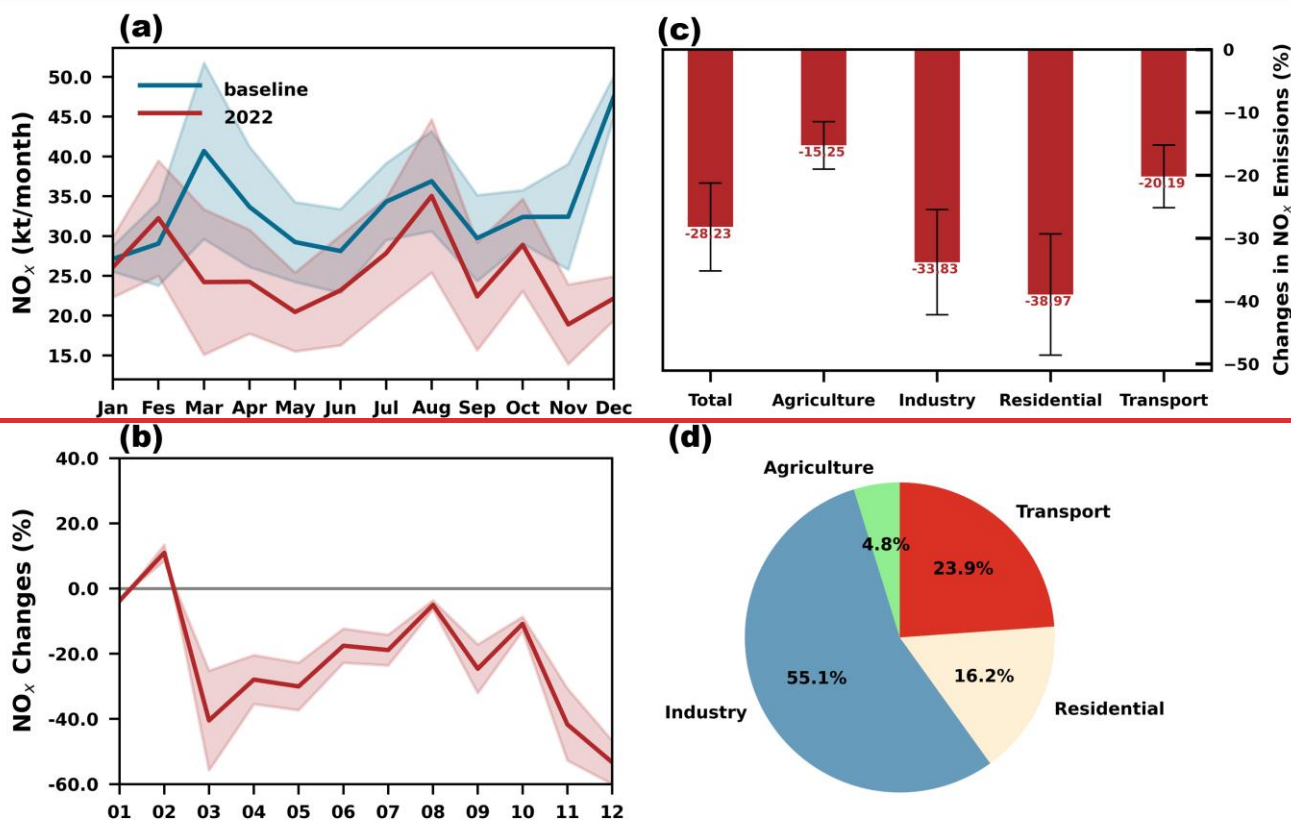


Fig. 24. Changes in anthropogenic NO_x emissions in Ukraine during 2022 and 2023 and their deviations from the baseline period. (a) Daily anthropogenic NO_x emissions in 2022 (red) and the baseline period (black). (b) Daily anthropogenic NO_x emissions in 2023 (blue) and the baseline period (black). (c) Relative differences in daily emissions in 2022 and 2023 compared to the corresponding days in the baseline period. (d) Sectoral contributions of prior (blue) and posterior (red) emissions in 2022 relative to the baseline. (e) Sectoral contributions of posterior emissions in 2023 (orange) relative to the baseline. (f) Contribution of each sector to the total reduction in emissions in 2022 (prior and posterior) and 2023, compared with the baseline year. ~~Changes in total NO_x emissions and in different sectors throughout Ukraine in 2022. (a) Monthly NO_x emissions in 2022 compared to the baseline. (b) Relative change in monthly emissions in 2022 compared to the baseline. (c) Relative change in emissions from different sectors during war compared to the baseline. (d) Contribution of different sectors to the decrease in emissions.~~

We examined the reductions in anthropogenic NO_x emissions across different sectors following the outbreak of war in 2022 based on both the prior and posterior estimates, as well as the sectoral changes in 2023 derived from the inversion (Fig. 2d, e), and the contribution of each sector to the total emission reduction (Fig. 2f). The inversion indicate that the industrial sector experienced the most significant impact from the war. Compared to the baseline, industrial emissions declined by 34% ($\pm 6.1\%$) in 2022 and by 24% ($\pm 4.4\%$) in 2023, accounting for 72% and 106% of the total annual reductions, respectively. These declines are comparable to those estimated from the prior inventory for 2022 and can largely be attributed to the heavy fighting and infrastructure disruption in the eastern industrial regions of Ukraine. Residential emissions also showed substantial reductions of 23% ($\pm 4.1\%$) in 2022 and 18% ($\pm 3.5\%$) in 2023. Despite contributing less to the total reduction due to their relatively

smaller share of emissions, the residential sector was the only sector in 2023 that did not exhibit a notable alleviation in its reduction rate. This persistence is likely associated with the population loss. The transport sector was a significant contributor to land-based anthropogenic NO_x emissions in Ukraine, but the observed decline was not as pronounced as that observed in the residential and industrial sectors. The reduction in transport emissions may have been partially offset by increased emissions from population displacement and logistical movements. Moreover, compared to the prior inventory, the inversion suggests a smaller reduction in transport emissions in 2022. This discrepancy could be due to the underestimation of military and emergency transport activities in energy-based inventories. In contrast to the prior inventory, which suggested a 3% decline in agricultural emissions in 2022, the inversion results indicate a 4% ($\pm 0.7\%$) increase in 2022 and a more pronounced 15% ($\pm 2.7\%$) increase in 2023. This discrepancy likely arises from the limitations of statistical data used in the prior inventory, which may have underestimated additional NO_x emissions from traditional farming practices and irregular land management under war conditions. The inversion results also suggest that the increase in agricultural emissions may partly reflect an overestimation of emissions in the central and western regions due to the assumption of fixed sectoral emission allocation.

~~We examined the reduction in sectoral NO_x emissions during the war relative to the baseline (Fig. 1c) and evaluated each sector's contribution to the overall emissions reduction (Fig. 1d). This war resulted in a reduction in emissions across all the sectors of Ukraine. The industrial sector experienced the most significant impact, accounting for 55% of the total decline, decreasing by 34% ($\pm 8.4\%$) compared with the baseline. This decline was primarily due to the impact of the ongoing war in eastern regions, where industry is well developed. The reduction in residential emissions was also notable, with a 39% ($\pm 9.6\%$) decline compared to the baseline, largely attributable to depopulation during the war. The transport sector was a significant contributor to land-based anthropogenic NO_x emissions in Ukraine, but the observed decline was not as pronounced as that observed in the residential and industrial sectors. The reduction in transport emissions was partially offset by emissions resulting from population movement and resource transfers. Moreover, the war affected eastern and southern Ukraine, which are the dominant wheat producers (Carriquiry et al., 2022), exhibited a considerable decline in agricultural emissions.~~

We compared the seasonal variations in sectoral emissions ~~between during the war period with those of in 2022 the and the~~ baseline (Fig. S2). Emissions across all sectors exhibited ~~increased or a~~ comparatively smaller decline during the summer months. ~~Agricultural emissions were significantly affected in the initial stages of the war (March–April) and during the cold season (November–December), with decreases of 26% ($\pm 9.6\%$) and 52% ($\pm 15.8\%$), respectively. By contrast, the war induced impact on NO_x emissions was less pronounced during the warm season ($1.4\% \pm 3.8\%$). T~~In 2022, the industrial emissions exhibited a notable decline of ~~39.38%~~ ($\pm 19.66.8\%$) in March, followed by a relatively stable period. A significant ~~70.17%~~ ($\pm 14.93.1\%$) decline was observed in emissions from the residential sector during the first three months of the war. Furthermore, a further decline in emissions occurred during the winter months owing to energy shortages. The transportation sector demonstrated notable responsiveness to increased transportation demand during the war, with a ~~28.24%~~ ($\pm 16.24.3\%$) decline in emissions observed in March, followed by a slight ~~decrease in April rebound~~. Nevertheless, ~~the decline in emissions remained at approximately 14% ($\pm 7.5\%$), o~~Wing to ~~military transport activities, humanitarian logistics, and evacuation-~~

related movement, the transport emissions was smaller than others in 8% ($\pm 1.4\%$). ~~the destruction of infrastructure and a shortage of drivers.~~ During the summer of 2023, agricultural emissions were approximately 6% higher than in 2022, indicating a gradual economic recovery in the central and western agricultural regions of Ukraine one year after the outbreak of the war. In contrast, the industrial sector exhibited substantial fluctuations in its emission reductions in 2023, likely reflecting repeated military operations in the eastern conflict zones. Transport-related NO_x emissions increased by 4.3% compared to 2022, indicating a gradual recovery in domestic mobility.

3.2 Decline in Different Regions

We examined the spatial distribution of total emissions (Fig. 2a3a, b) and ~~monthly-daily~~ NO_x emissions (Fig. S3, 4) during the war relative to the baseline. Most regions of Ukraine exhibited a reduction in NO_x emissions of approximately ~~30~~15% (Fig. 2a3a) ~~during the war in 2022~~. This reduction was more pronounced in eastern regions of Ukraine than in the central and western regions. A slight increase in emission was observed in parts of the ~~Kyiv~~, Zhytomyr, Rivne, and ~~Sumy-Khmelnytskyi~~ oblasts (Fig. 2a3a). Increases in the NO_2 VCDs in these regions were observed using the TROPOMI satellite (Fig. S41a), which were mainly from the agricultural and transportation sectors (Fig. S5S5), possibly because these regions are located ~~afar~~ away from the front of the war around the Ukrainian capital, ~~and Eastern Ukraine~~ where a larger population has gathered. After April 2022, the primary theater of the Russia–Ukraine war was concentrated in the eastern oblasts bordering Russia. The central and western regions exhibited a weaker impact than the eastern regions. In terms of the spatial distribution of sectoral variations (Fig. S5S5), the most pronounced decline in agricultural emissions was observed in eastern and southern Ukraine. This may be attributed to the region's importance as a winter wheat-producing area (Lin et al., 2023) and the disruption of cultivation caused by the Russian army's control, which caused a reduction in fertilizer application and agricultural NO_x emissions. Spatially, industrial emissions exhibited a notable decline throughout Ukraine, particularly in Luhansk Oblast. The war had a direct or indirect impact on industrial production across Ukraine, with the most pronounced damage observed in conflict zones. ~~The outbreak of the war during this period resulted in the destruction of numerous Ukrainian residential facilities and the displacement of numerous refugees to the central and western regions and abroad. Consequently, emissions increased as the war stabilized in the eastern regions, housing conditions improved, and some refugees returned home.~~ ~~The outbreak of the war during this period led to widespread destruction of residential infrastructure and the mass displacement of civilians. Large numbers of refugees relocated to central and western Ukraine or fled abroad, resulting in a sharp decline in residential NO_x emissions. In 2023, NO_x emissions in central and western Ukraine increased compared to the baseline, while emissions in the eastern regions remained suppressed due to ongoing localized conflict (Fig. 3b). In 2022, NO_x emissions in Crimea showed a slight decrease in the western region and a modest increase in the east, primarily due to the concentration of Russian military logistics and air force operations in the eastern part of the peninsula. In 2023, emissions increased further across the region, reflecting Crimea's growing role as a long-term base for Russian troop deployment and logistical support. Sectoral analysis~~

reveals varying degrees of recovery in agriculture, industry, and transportation across central and western Ukraine. In contrast, residential emissions in 2023 remained at levels comparable to those in 2022, consistent with the national trend. These findings suggest that while efforts to reestablish agricultural and industrial activity have taken hold in the rear regions of the battlefield, residential activity has yet to recover, likely due to continued population displacement.

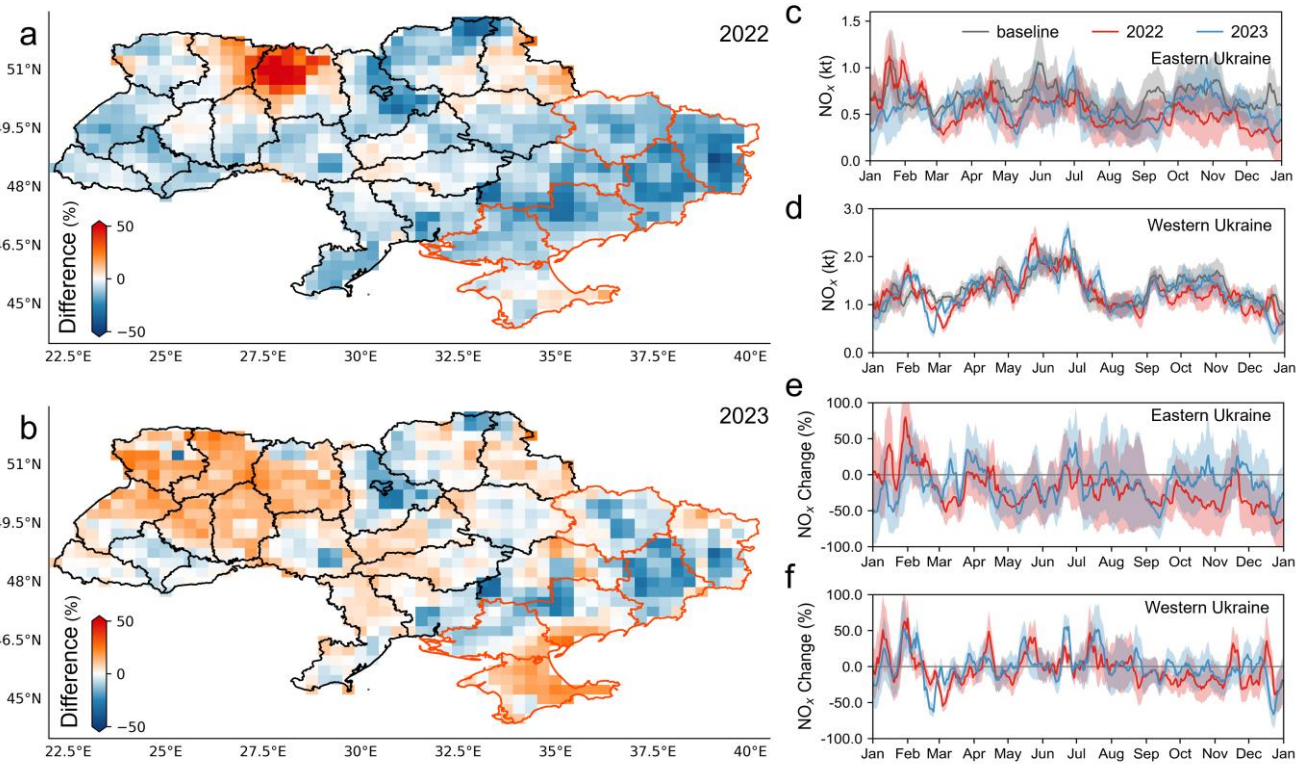


Fig. 3. NO_x emissions changes in different regions. (a~b) Spatial distributions of NO_x emissions changes in Ukraine during war in (a) 2022 and (b) 2023 relative to the baseline. Eastern Ukraine is marked with red lines, and Western Ukraine is in black. (c~d) Daily NO_x emissions in 2022, 2023 and the baseline years in (c) Eastern Ukraine and (d) Western Ukraine. (e~f) Relative changes in daily NO_x emissions in 2022 and 2023 relative to the baseline in, (e) Eastern Ukraine, and (f) Western Ukraine.

To assess the geographical difference in NO_x emissions during the war in Ukraine, we categorized Ukraine into Eastern Ukraine and Western Ukraine based on the scope of the war and analyzed the seasonal variations in satellite NO₂ VCDs in both regions (Fig. 2a3a, b). Subsequently, we calculated the seasonal fluctuations in anthropogenic NO_x emissions in comparison with the baseline for different regions (Fig. 2-3b~ef). During the 2022 wartime period, anthropogenic NO_x emissions in Eastern and western Ukraine decreased by 28.8% ($\pm 5.2\%$) and 8.4% ($\pm 1.5\%$), respectively. Both the Eastern Ukraine and Western Ukraine emissions exhibited a notable decline in March, with reductions of 4030.9% ($\pm 155.6\%$) and 4125.8% ($\pm 23.74.6\%$), respectively, caused by a large-scale attack that began in late February 2022. Subsequently, Western Ukraine emissions exhibited a gradual recovery from April to August, whereas Eastern Ukraine emissions remained low in the subsequent months,

reaching their lowest (~~4632%~~~~±115.9%~~) in May. This discrepancy may be attributed to the initial direct impact of the war on the Western Ukraine in March, which resulted in a rapid decline in emissions. The Eastern Ukraine, with its smaller population and dominant industrial sector, did not minimize emissions in the first month. However, as the war continued, emissions continued to decline due to population loss, energy shortages, and infrastructure damage. In both regions, the decline in emissions intensified concurrently after November, likely because of the inability of energy shortages to meet increased demand during the winter months.

In 2023, anthropogenic NO_x emissions in Ukraine also exhibited a marked spatial divergence between the eastern and western regions. The Eastern Ukraine experienced a sustained emission decline of 17.0 % (±3.1%) compared to the baseline, largely attributed to industrial inactivity under Russian control, destruction of power infrastructure, and ongoing population displacement. In contrast, emissions in the Western Ukraine declined by only 2.9 % (±0.5%), reflecting greater resilience due to the westward relocation of industry, influx of international assistance, and support from adaptive agricultural practices.

Seasonal trends reveal that eastern emissions reached their lowest points in January (-35.8 %±6.4%) due to exacerbated energy shortages, in April (-16.2 %±2.9%) during intensified military offensives, and in September (-31.7 %±5.7%) following a second collapse of the power grid. The smallest reduction (11.9 %±2.1%) was observed in June, likely driven by intensified military activity during Ukraine's counteroffensive. In the west, the most substantial reduction occurred in February (10.0%±1.8%) and December (22.9%±4.1%), primarily due to gas supply disruptions by Russia and regional power grid failures. Emissions in other months remained comparable to or slightly above baseline levels.

~~A comparison of sectoral emissions in the Eastern Ukraine and Western Ukraine (Fig. S6) revealed that the industrial and residential sectors were the most significantly impacted emission sectors in both regions. This underscores the emergence of energy shortages as a major challenge for Ukraine amid the ongoing war. Due to the geographical ramifications of the war, all sectors in Eastern Ukraine experienced more pronounced declines in emissions than Western Ukraine. In Western Ukraine, except for the industrial and residential sectors, which exhibited reductions of 32% (±7.9%) and 36% (±9.0%), respectively, the other sectors exhibited reductions not exceeding 20% (±1.5%). Nevertheless, the agricultural and transport sectors contributed more to the overall decline in emissions in Western Ukraine than they did in Eastern Ukraine. This was primarily attributed to the region's more advanced agriculture and larger population.~~

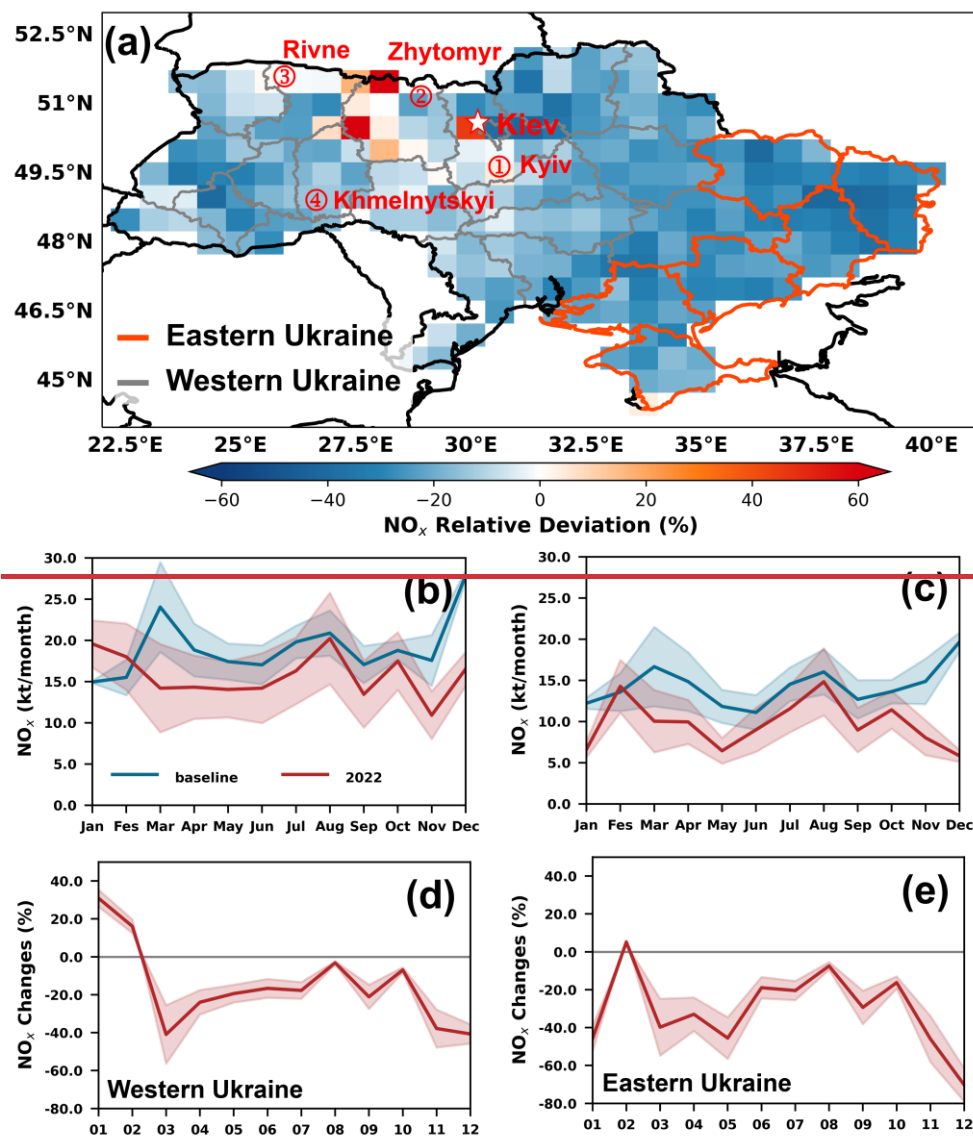


Fig. 2. NO_x emissions changes in different regions. (a) Spatial distributions of NO_x emissions changes in Ukraine during war relative to the baseline. Eastern Ukraine is marked with red lines, and Western Ukraine is in grey. Cities with increased emissions and the capital city are also marked. (b) Monthly NO_x emissions in 2022 and the baseline years in Western Ukraine, (c) monthly emissions in Eastern Ukraine, (d) relative changes in monthly NO_x emissions in 2022 relative to the baseline in Western Ukraine, and (e) relative changes in Eastern Ukraine.

3.3.3 Comparison With the COVID-19 Pandemic

To provide a more comprehensive representation of the impact of the war in Ukraine, we further analyzed changes in NO_x emissions during the 2020 COVID-19 pandemic. NO_x emissions in Ukraine during the pandemic (~~April~~May–December) in 2020 decreased by ~~133.9%~~ (~~±4.10.6%~~) compared with those in the baseline year. After falling by ~~398.5%~~ (~~±12.81.5%~~) for the first time in ~~April~~May, NO_x emissions in Ukraine in 2020 returned to levels comparable to the baseline ~~in May~~, with a decline of only ~~32%~~ (~~±0.84%~~) (Fig. ~~S7S6~~). ~~The spatial distribution of anthropogenic NO_x emissions in Ukraine during the pandemic was more homogenous, with no significant east-west differences.~~ During the pandemic, the main declining sectors were ~~agriculture (-0.8%±0.1%)~~industry and transportation (~~15%±3.8-2.6%±0.5%~~). ~~The agricultural and residential sectors displayed an upward variation in most regions, whereas decreases primarily occurred in the Kiev periphery and southeast regions (Figs. S4, S8).~~ The changes in anthropogenic NO_x emissions during the pandemic are described in detail in Text S1 of the Supplementary Material.

The decline in NO_x emissions owing to the war was ~~substantially exceeded twice as large as~~ the impact of the pandemic. ~~Throughout most months after May 2020, NO_x emission levels remained comparable to the baseline year. This was followed by further declines in the individual months after May,~~ which were less ~~reduction~~ pronounced than those observed in 2022. ~~The emission decline was primarily concentrated in the transport sector, fundamentally driven by mobility restrictions imposed under lockdown measures. Sectoral emissions were not affected to the same extent during the pandemic as during the war. The industrial and transport sectors exhibited a general decline, primarily because of the implementation of home quarantine policies, which resulted in a reduction in residential mobility and industrial production.~~ This comparison highlights the more profound and enduring impact of the war on Ukraine compared to the pandemic. The pandemic slowed national human activity without causing far-reaching damage.

4. Discussion

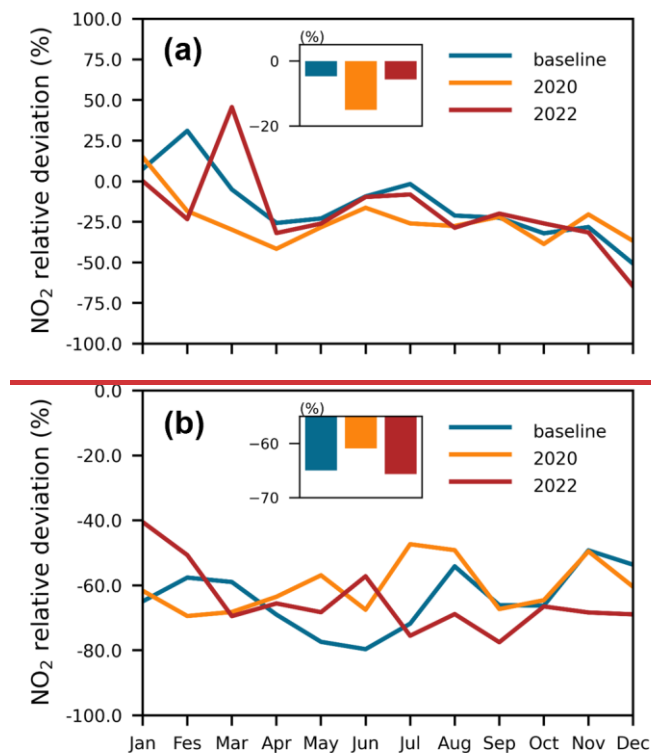
Our study inverts changes in anthropogenic NO_x emissions during the Russia–Ukraine war in 2022 ~~and 2023~~ based on satellite observations. We explore the capacity of satellite-based NO_x emission inventories to monitor economic production activities within Ukraine-affected regions amid the backdrop of frequent localized conflicts.

4.1 Evaluation

We conducted a comprehensive validation to ensure the reliability of the inverted NO_x emissions in Ukraine. Compared with the satellite-observations, the accuracy of NO₂ VCDs simulated using posterior emissions significantly improved; the ~~average discrepancies~~RMSE between the prior and posterior simulated NO₂ VCDs were ~~1.320.57~~ Pmolec/cm² and ~~0.1340~~ Pmolec/cm², respectively (Figs. ~~S9, S107~~). The posterior emissions more accurately reflect the seasonal fluctuations in NO₂ VCDs in Ukraine (Fig. ~~S9S8~~). Prior ~~simulated NO₂ VCDsemissions~~ exhibited an ~~overestimation-underestimation~~ of ~~80.311.6%~~ at the

national level, with the greatest discrepancy observed in the ~~Central-southwest~~ and ~~Southern-northeast~~ regions. Furthermore, we used surface in-situ observations over Europe to assess the accuracy of the simulated NO₂ concentrations with prior and posterior emissions. The simulated concentrations with posterior emissions exhibited higher R² and lower RMSE values at most sites across Europe (Fig. ~~S11-S9~~a, b). Furthermore, the simulated values of a posterior emissions exhibited greater accuracy in fitting between sites than the prior emissions (Fig. ~~S11-S9~~c).

Despite the improvement over the prior inventory, some discrepancies between the posterior simulations and both satellite and surface in-suit observations remain. To assess the impact of these discrepancies on the interannual variability of the results, we calculated the relative deviation between the simulated and observed values across different years, and computed the seasonal variations in the relative deviations on a daily (with satellite) or monthly (with surface in-situ) basis (Fig. ~~3S10~~). The results indicate that the relative differences between the posterior simulations and satellite observations remained within ±50 % on more than 95 % of the days. The results show that the relative deviation between the posterior emission simulations and satellite observations remained within ±50% across these years. The largest fluctuations in the relative deviation were observed between January and March, with deviations stabilizing for the remaining months. In contrast, when comparing with surface ~~in-situ~~in-suit observations, the relative deviations were concentrated between -40% and -~~80~~70%. Although these results did not exhibit the same seasonal fluctuations observed in the satellite comparison, they demonstrated lower overall variability compared to the satellite observations. Given that the relative deviations between the posterior simulated concentrations and satellite observations remained within a consistent range across different years, we can conclude that the inverse ~~sted~~ on NO_x ~~emission changes-results~~ maintain high consistency across the years.



4.2 Comparison with other results

Fig. 3. Seasonal distribution of the relative deviations in posterior simulations in the baseline, 2020 and 2022, with respect to (a) satellite observations and (b) surface in-suit observations. The bar subplot shows the mean annual relative deviation.

Additionally, according to the World Bank's Ukraine Rapid Damage and Needs Assessment (the World Bank, 2023), the war resulted in a 29.2% decline in Ukraine's GDP for 2022, which closely aligns with our estimated 28% reduction in emissions. We collected energy consumption data, industrial production index (IPI) data, the gross domestic product (GDP) and agricultural GDP from the State Statistics Committee of Ukraine (SSCU), and CO₂ emissions data from the Global Carbon Project (GCP) (Friedlingstein et al., 2023; Jones et al., 2021) (Table 1). The SSCU and GCP provide data only for different energy types. For comparison, we assumed that oil was consumed primarily in transportation, natural gas was used primarily by residents, and coal was mainly consumed by industry. Compared with that at baseline, the inverted NO_x emissions in 2022 declined by 14%, 28%, and 32% in the transportation, residential, and industrial sectors, respectively. The relative changes in the inverted NO_x emissions of each sector are highly consistent with the relative decline in oil and gas consumption data from the SSCU. Ukraine's agricultural sector saw the smallest change in emissions in 2022, with a decline of 13%. This result is corroborated by the relatively small changes in agricultural GDP volumes reported by the SSCU. In addition, the rate of decrease in NO_x emissions in the transport and residential sectors was close to the relative changes in the CO₂ emissions of oil and gas from the GCP, respectively. However, the CO₂ emissions in Ukraine in 2022 declined by 35% compared with those at baseline, which was approximately 10% more than the inverted anthropogenic NO_x emissions. According to the results in Table 1, this difference is partly due to the decline in emissions from coal combustion, which exceeds the IPI data from SSCU and

550 ~~the industrial sectors emissions from our inversion by about 10%. This difference may be due to the reduced utilization of coal in industrial production.~~

According to the World Bank's Ukraine Rapid Damage and Needs Assessment (the World Bank, 2023), Ukraine's GDP declined by 29.2% in 2022 and 26% in 2023. This drop is consistent with the 23% reduction in anthropogenic NO_x emissions estimated by the bottom-up EDGAR inventory (prior emissions) but significantly exceeds the 15% and 8% decrease in 2022 and 2023 inferred from TROPOMI observations in this study. Further analysis using SSCU data revealed that oil and natural gas consumption in Ukraine decreased by 15% and 34%, respectively, in 2022, and by 13% and 32% in 2023 (Table 1). Given that oil consumption is primarily associated with the transportation sector and natural gas consumption predominantly reflects residential activities, the observed reductions in oil consumption align well with the bottom-up estimates for transport emissions but are larger than the corresponding decreases captured by our satellite-based inversion. In contrast, the decline in natural gas use far exceeds the reduction in total anthropogenic NO_x emissions from both approaches. Agricultural GDP dropped by 5% in 2022 and increased by 5% in 2023, while satellite-derived emissions indicated a stronger recovery in agricultural NO_x emissions. For the industrial sector, the Industrial Production Index (IPI) declined by 38% in 2022 and 33% in 2023, which is largely consistent with the emission reductions inferred from our inversion framework.

In addition, we compared the changes in anthropogenic NO_x emissions with fossil fuel CO₂ emissions reported by the Global Carbon Project (GCP) (Friedlingstein et al., 2023; Jones et al., 2021) (Table 1). The GCP inventory attributes CO₂ emissions by fuel type. Consistent with previous assumptions, we considered oil and natural gas to primarily reflect emissions from the transport and residential sectors, respectively, while coal combustion is assumed to dominate industrial emissions. The results show that GCP-reported emissions from oil, natural gas, and coal declined by 14%, 32%, and 46%, respectively, in 2022, and by 13%, 30.4%, and 51.3% in 2023. The reductions in oil and gas-related CO₂ emissions are consistent with national energy consumption statistics, reflecting the common bottom-up data sources used in both GCP and national energy reporting. However, the decline in coal-related emissions reported by GCP is substantially larger than that indicated by both the inversion-derived NO_x emissions and energy statistics. This discrepancy may be attributed to a reduction in coal utilization efficiency or structural changes in industrial energy use during the war period.

The closer alignment between energy statistics and the bottom-up EDGAR and GCP inventory is expected, as EDGAR and GCP rely directly on national statistical reports for activity data, resulting in high consistency with macroeconomic trends. In contrast, the satellite-constrained inversion incorporates spatial heterogeneity and seasonal variability, enabling pixel-level adjustments and rebalancing of sectoral contributions based on EDGAR's original distribution. This leads to systematic deviations in sector-level trends from national statistics. Additionally, discrepancies between the inversion results and energy statistics may stem from the satellite's ability to capture unconventional sources of NO_x emissions during wartime—such as those from military vehicles and artillery. Moreover, widespread damage to conventional energy infrastructure (e.g., electrical grids) necessitated the use of temporary power generation (e.g., diesel generators), which is likely not reflected in official energy consumption reports (Wang et al., 2024). Therefore, under the extraordinary conditions of armed conflict, satellite-

constrained inversions offer a valuable complement to traditional statistical assessments by capturing both conventional and conflict-induced NO_x emissions.

585 **Table 1. Changes in GCP CO₂ emissions and SSCU statistics for baseline, ~~2020 and 2022~~ and 2023, and relative changes in ~~2020 and 2022~~ and 2023 compared with the baseline.**

	Sector	Transport (kt)	Residential (kt)	Industry (kt)	Agriculture (kt)
	Baseline	182.1	32.6	209	73.6
	2022	154.8/-15%	23.6/-28%	131.3/-37%	71.8/-3%
	2023	-	-	-	-
Inverted NO _x emissions	Sector	Transport (kt)	Residential (kt)	Industry (kt)	Agriculture (kt)
	Baseline	313.2	48.2	246	125.8
	2022	295.8/-6%	39.1/-19%	176.1/-28%	135.4/8%
	2023	307.8/-2%	37.9/-21%	174.5/-29%	144.4/15%
SSCU*	Sector	Oli (tb/d)	Gas (b Nm ³)	IPI	Agricultural GDP (Mh)
	Baseline	234.5	2.68	102.3	474965
	2022	200.5/-15%	1.90/-29%	63.9/-38%	449148/-5%
	2023	203.8/-13%	1.81/-33%	68.2/-33%	500540/5%
GCP CO ₂	Sector	Oli (Mt)	Gas (Mt)	Coal (Mt)	-
	Baseline	37.1	59.4	106	-
	2022	31.7/-14%	40.3/-32%	57.6/-45.6%	-
	2023	32.2/-13%	41.3/-30%	51.5/-51.3%	-

Inverted NO _x emissions			SSCU*			GCP CO ₂		
Sector	Baseline	2022	Data type	Baseline	2022	Data type	Baseline	2022
Transport (kt)	131.2	113.4/-14%	Oli (tb/d)	234.5	200.00/-15%	Oli (Mt)	234.5	200.00/-15%
Residential (kt)	48.3	34.7/-28%	Gas (b Nm ³)	2.65	1.90/-28%	Gas (Mt)	2.65	1.90/-28%
Industry (kt)	187.5	128.4/-32%	IPI	102.3	63.86/-38%	Coal (Mt)	102.3	63.86/-38%
Agriculture (kt)	34.2	29.6/-13%	Agricultural GDP (Mh)	474,965	449,148/-5%	-	-	-

* Oil data of SSCU in thousand barrels per day, Gas data of SSCU in billion Nm³, IPI calculated using ~~2006-2016~~ as 100. Agricultural GDP in Ukraine in Millions of Ukrainian hryvnias.

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4.3 War-related NO_x emissions

As mentioned above, War itself may also lead to large amounts of NO_x emissions, including military vehicles and artillery. Thus, we further assessed the war-related NO_x emissions. We firstly applied Locally Estimated Scatterplot Smoothing (LOESS) and residual analysis to quantify the spatial heterogeneity of significant emission anomalies in the 2022 wartime period and 2023 (Fig. 4a, b). Significant emission anomalies were identified by calculating the cumulative significant residuals (exceeded

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three times the standard deviation) of daily grid-level emissions relative to the corresponding 30-day LOESS-smoothed values. The results indicate localized positive residual anomalies (red hotspots) were observed along conflict frontlines and logistical hubs, suggesting that military operations and emergency responses significantly elevated emissions in these areas. This is consistent with the severely damaged areas identified by Priyanka Gupta et al. (Gupta and Shukla, 2024) using NASA MODIS FIRMS active fire detections. We analyzed daily anthropogenic NO_x emissions alongside verified reports of military activity in Ukraine (2022–2023) from BBC and Reuters, and found strong correspondences between emission anomalies and key military events. In Kyiv and Kharkiv, emissions sharply declined following the outbreak of war on 24 February 2022, reaching minima during periods of civilian shutdown. Emissions rebounded during March as military logistics and emergency operations intensified. Kharkiv and Luhansk showed short-term positive anomalies during Ukrainian counteroffensives and Russian reinforcements, while Donetsk experienced sustained negative anomalies due to prolonged conflict and infrastructure damage. Crimea, as a logistical hub, showed persistent emission increases linked to military operations. These findings highlight the potential of high-frequency NO_x emissions as a proxy for monitoring the intensity and evolution of wartime activities.

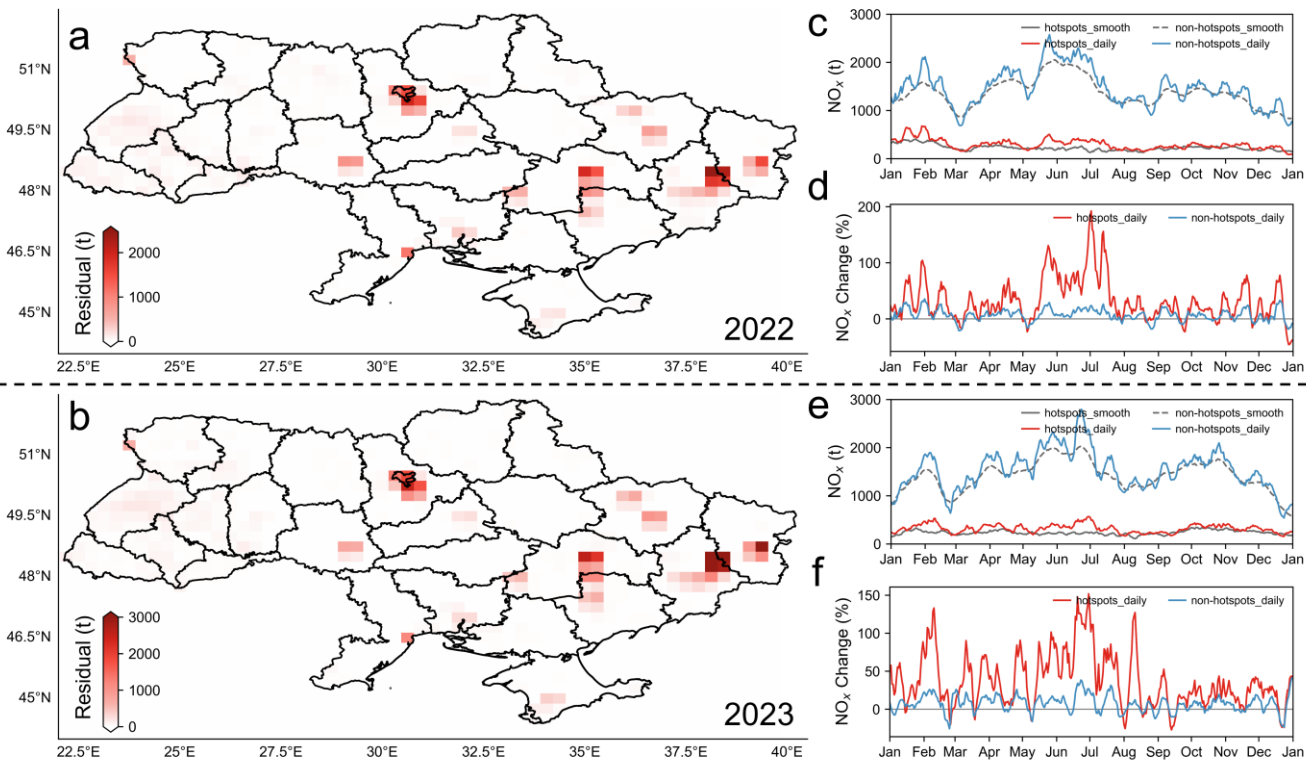


Fig.4 Spatiotemporal distribution of significant daily anthropogenic NO_x emission anomalies in Ukraine during the 2022 war period (after 24 February) and 2023. (a, b) Spatial distribution of daily cumulative significant emission anomalies relative to the LOESS-smoothed emissions in 2022 (a) and 2023 (b). (c, d) Seasonal variation of daily emissions and LOESS-smoothed emissions (c), and corresponding

relative differences (d), for 2022 in war-affected hotspots (grid cells with daily anomalies $> 3\sigma$) and non-hotspot regions. (e, f) Same as (c, d), but for 2023.

The emission anomalies observed in the inversion results suggest that wartime activities made a non-negligible contribution to overall NO_x emissions. By comparing daily emissions with their corresponding LOESS-smoothed values, we classified each day's spatial emissions into war-affected hotspots (grid cells with daily anomalies $> 3\sigma$) and non-hotspot regions. Results show that during the war periods of 2022 and 2023, the smoothed emissions in hotspot areas were 30.8% and 35.6% lower than the corresponding daily emissions, respectively. In contrast, differences in non-hotspot regions were only 6.7% and 8.2%. This indicates that smoothing effectively filtered out the high-frequency variability associated with military activities in hotspot areas. With this method, we further estimated the relative deviation of the smoothed emissions from the pre-war baseline, finding reductions of 23% in 2022 and 18% in 2023, which closely matching the emission decreases reported by the EDGAR inventory. This suggests that military-related activities offset approximately 8% and 10% of the overall emission reductions in 2022 and 2023, respectively, which partially explains the lower reductions in our inversion than those from bottom-up inventories and independent economic data. However, quantifying the exact compensatory effect of direct war emissions on emission reductions remains methodologically challenging. Because these sources are inherently episodic and spatially concentrated, complicating their separation from background variability in sectoral inventories. These findings highlight warfare as a distinct emission modulator that can temporarily reshape regional source profiles, though its aggregate contribution remains secondary to economy-wide suppression effects in determining net emission trajectories.

4.4 Uncertainty analysis

The inversion emissions increased by 39.9% ($\pm 7.2\%$) compared to the prior inventory, which is within the 50% uncertainty range of EDGAR (Crippa et al., 2018). The primary sources of uncertainty in our inversion framework can be attributed to two main factors: the quality of satellite observations and the inherent limitations of the inversion process. To enhance data reliability, we utilized TROPOMI NO_2 VCD observations processed using a more accurate processor version. Despite this improvement, TROPOMI observations still substantially underestimate tropospheric NO_2 levels (East et al., 2022). To quantify the uncertainty introduced by satellite NO_2 VCD observations, we estimate that ~~these~~ uncertainties of observations contribute an average of $\pm 6.56.7\%$ uncertainty to the inversion results.

Due to the fixed overpass time of TROPOMI at approximately 13:30 local time, the inversion is constrained to reflect emissions around this midday window, limiting its sensitivity to nocturnal emission variations. This temporal sampling introduces a limitation in capturing potential shifts in emission timing, particularly under wartime conditions. During the Russia–Ukraine war, especially in high-risk zones, there may have been a redistribution of human and industrial activities toward nighttime hours due to safety concerns, power rationing, or tactical considerations such as avoiding aerial detection. While such behavioral shifts could potentially increase nocturnal emissions, the current inversion system is unable to capture these changes due to the absence of nighttime satellite data. Thus, while our results provide a robust estimate of daytime anthropogenic NO_x emissions, they may overestimate total emission changes if substantial nocturnal activity occurred. Future work integrating

ground-based measurements, high-temporal-resolution modeling, or geostationary satellite data (e.g., TEMPO, Sentinel-4) could help address this temporal limitation.

-Further, the inversion framework itself introduces uncertainty, which stems from both the missing data of satellite, particularly during the winter months at high latitudes such as in Ukraine, and the process used to derive emissions, and we quantified this part of the uncertainty using the OSSE. The OSSE results show that the inversion framework effectively reduces the uncertainty in the prior and maintains high spatiotemporal consistency with the assumed true values (Fig. ~~S12S11~~). The bias between the posterior emissions and the true values was ~~0.693.9%~~, a ~~98.2788.3%~~ reduction from the bias in the prior. Additionally, the posterior uncertainty without considering natural sources was ~~1.08.6.4%~~, reducing the prior uncertainty by ~~97.2480.8%~~. Seasonally, by accounting for natural source emissions, the inversion uncertainty for anthropogenic NO_x emissions decreased ~~2.54.03%~~ for the year-round, with the largest improvement observed from ~~June-May~~ to ~~October-August~~ when natural emissions are higher, with an average improvement of ~~18.251.4%~~ ~~in July~~. However, the OSSE also revealed that despite the model's effectiveness in reducing prior error, ~~2.311.7%~~ of uncertainty remained. Thus, after combining the uncertainties from both sources, the total uncertainty in the reduction of emissions during the war period in the inversion results is estimated to be approximately ~~7.618.4%~~. The uncertainties across different stages, sectors, and regions are provided in the results. In addition, the inversion was conducted at a spatial resolution of 0.25° × 0.3125°, which may smooth or omit localized emission signals, leading to potential biases in the estimated NO_x emissions.

5. Conclusion

Our results indicate a notable reduction in NO_x emissions during the war, highlighting the severe impact of the war on the Ukraine's socioeconomic activities. Sector-specific analysis revealed the most significant declines in the industrial and residential sectors. And the eastern Ukraine experiencing more severe disruptions. Additionally, the energy shortages during winter exacerbated the decline in emissions, illustrating the compounded effects of infrastructure damage and resource scarcity. The stalemated war has led to substantial declines in Ukraine's energy and industrial production levels, causing significant damage to housing and transportation infrastructure. ~~Rural areas and agricultural production were also significantly affected, causing further economic destabilization. This disruption in both urban and rural areas resulted in population displacement towards safer regions and even other countries, exacerbating labor shortages and further stalling economic activities.~~ By capturing daily emission anomalies, NO_x serves as an effective indicator of fluctuations in emissions caused by war-related destruction and military activities, providing a direct and timely metric for monitoring armed conflict dynamics. In comparison, the 2020 COVID-19 pandemic resulted in less severe and shorter-term reductions in NO_x emissions, primarily affecting industry and transportation due to home quarantine policies. This reduction was temporary and primarily driven by public health measures that aimed to protect lives while allowing for a relatively quick economic rebound post-lockdown. Conversely, the destruction during the war of industrial capacity, residential areas, and transportation infrastructure, coupled with energy shortages, has not only disrupted current activities but has also compromised future development potential.

This study emphasizes the interconnectedness of environmental sustainability and socio-economic stability. Changes in NO_x emissions can only partially quantify the economic impacts of the war, and the actual socioeconomic impacts of the war are likely to be far greater than estimates derived from satellite observations. While our analysis is only a snapshot of the impacts of war, the findings have far-reaching implications for further research and for informing post-war reconstruction planning, energy security strategies, and emission mitigation policies. ~~While our analysis is only a snapshot of the impacts of war, the findings have far-reaching implications for further research and policy development.~~

Data availability

The TROPOMI NO₂ VCDs data are available at <https://www.temis.nl/> ~~https://seihub.eoernicus.eu/~~, the full versions of the GEOS-Chem model and the driver data are available at <https://geos-chem.readthedocs.io/en/latest/>. Data of energy consumption and IPI are from the Statistical Office of Ukraine at <https://www.ukrstat.gov.ua/>. CO₂ emissions of GCP (GCP-GridFEDv2023.1) are available at <https://zenodo.org/records/8386803/>. The inversion of anthropogenic NO_x emissions in this paper are available at <https://zenodo.org/records/12540012>.

Author contribution

Y.M. and F.J. conceptualized the study. Y.M., H.M.W., and F.J. developed the methodology. Y.M., S.Z.F., and M.W.J. conducted the investigation. Y.M. performed the visualization. W.M.J. and F.J. supervised the project. The original draft was written by Y.M., and L.Y.L., H.K.W., and W.M.J. contributed to the review.

Competing interests

The authors declare that they have no conflict of interest.

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