

Supplementary Materials for

Satellite-based estimates of decline and rebound in China's CO₂ emissions during COVID-19 pandemic

Bo Zheng, Guannan Geng, Philippe Ciais, Steven J. Davis, Randall V. Martin, Jun Meng, Nana Wu, Frederic Chevallier, Gregoire Broquet, Folkert Boersma, Ronald van der A, Jintai Lin, Dabo Guan, Yu Lei, Kebin He, Qiang Zhang*

*Corresponding author. Email: qiangzhang@tsinghua.edu.cn

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Supplementary Materials

MEIC emission model

The MEIC emission model was developed by Tsinghua University (28, 42). MEIC tracks the evolution of the manufacturing and pollution control technologies for more than 700 anthropogenic emission sources in China, which can be integrated into five source sectors of power, industry, residential, transportation, and agriculture. The coal-fired power plants in the power sector (43, 44) and the cement plants in the industry sector (45), the two important sources of NO_x and CO₂ emissions in China, are treated as point sources with accurate geographic locations, where emissions are estimated using the facility-level activity data and emission factors. The other emission sources are all estimated as area sources using the province- or county-level parameters (46, 47). The MEIC emission inventory has an accurate representation of emission spatial distributions (48, 49) and emission annual trends (41, 50), evaluated by satellite observations. The MEIC model uses monthly and daily temporal profiles including electricity generation, industry factory operating rate, traffic volume index, and heating degree day (i.e., the number of degrees that a day's average temperature is below 18 °C) to disaggregate the annual emission estimates to monthly and daily emissions. The daily sectoral NO_x emissions from January to April in 2019 are shown in Figure S2a. The low values of CO₂ to NO_x emission ratio around the Chinese New Year in 2019 (Figure S2b) are due to reduced activities in power and industry sectors during the holiday.

Sensitivity simulations of β value

The estimation of β has been proved robust at coarse spatial resolutions. Lamsal et al. (29) used the global GEOS-Chem model ($2^\circ \times 2.5^\circ$) to perform several tests of β . They find that a perturbation of 30% NO_x emissions instead of 15% changes global averaged β by <2%. Increasing anthropogenic VOCs and CO by 15% increases the global value of β by 2.8% and 1.0%, respectively. When NO_x emissions are only perturbed in a single grid cell in Ohio, β values in neighboring grids are affected by 2–6%, which indicates that β is barely influenced by the transport of NO₂ from neighboring grid cells at the resolution of $2^\circ \times 2.5^\circ$.

In this study, we use the nested-grid model with a higher spatial resolution ($0.5^\circ \times 0.625^\circ$) than the global model used in Lamsal et al. (29), which resolves the nonlinear NO_x chemistry and heterogeneous emission sources. However, β obtained from smaller grid cells might be more affected by the transport of NO₂ from neighboring grids or background area, especially during winter time when the NO₂ lifetime is longer. We conduct a sensitivity test using the global GEOS-Chem model to simulate β as a comparison. Results show that the national average value of β is similar to the nested-grid model, and the estimated NO_x emissions are within a 3% difference.

We also find that perturbing NO_x emissions by 30% or 50% instead of 40% only changes β by -0.7% and 0.8%, respectively. We also conduct sensitivity simulations with an additional 20% decrease in CO and a 40% decrease in VOCs, which increase β by 2.5% nationally. The results of the sensitivity tests show that the β value is not sensitive to emission perturbation.

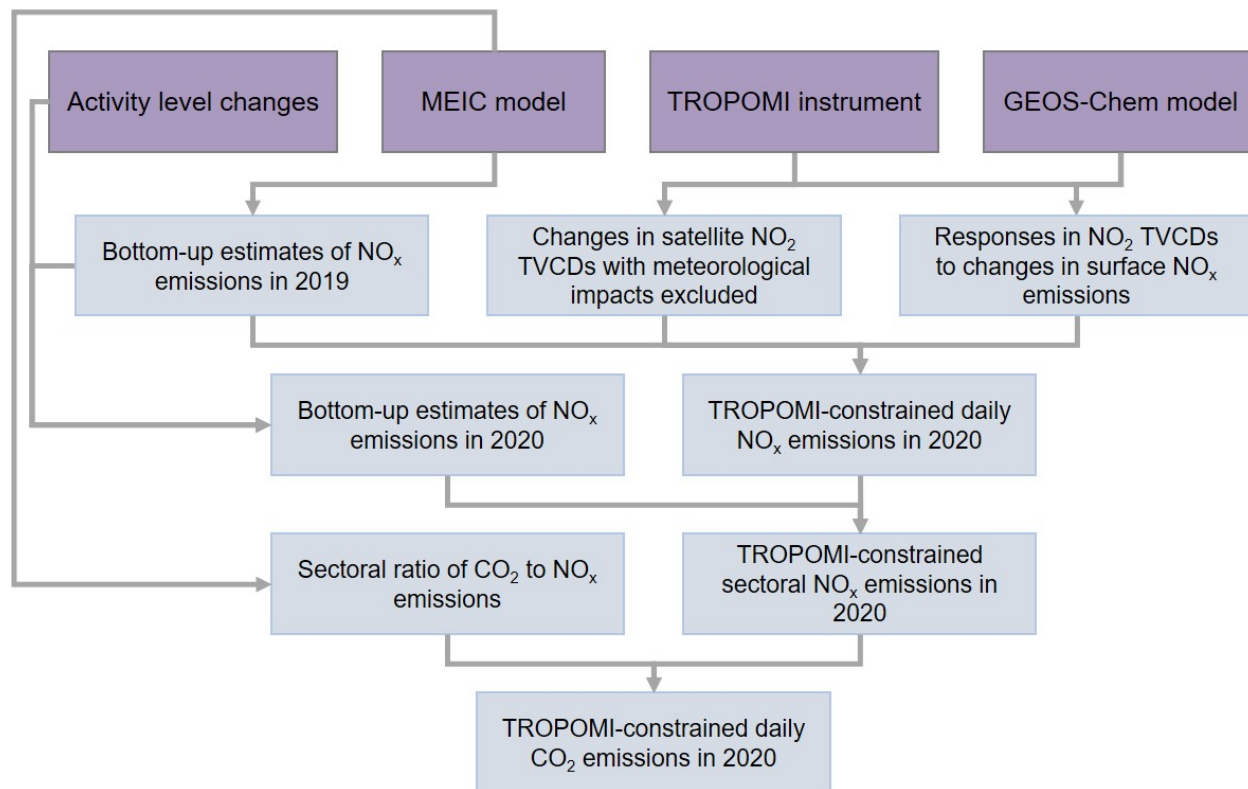


Fig. S1. Methodology framework to infer daily CO₂ emissions from TROPOMI NO₂ TVCDs over China.

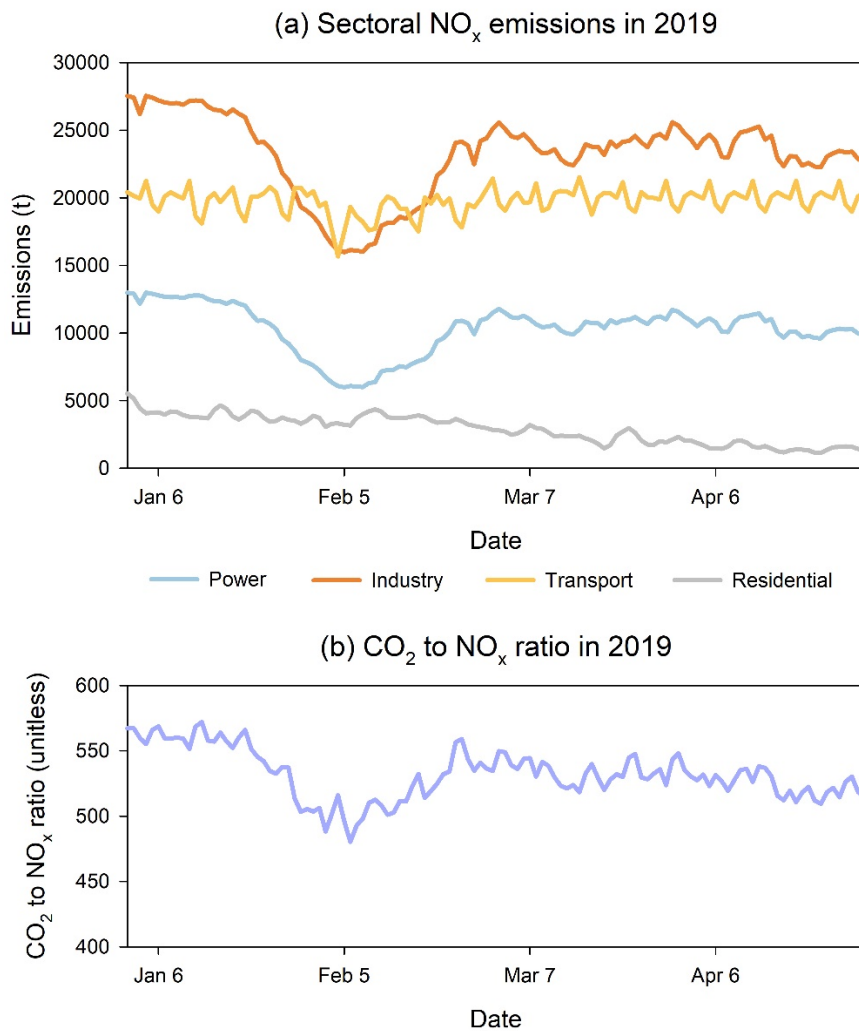


Fig. S2. MEIC daily emissions from January to April in 2019. Fig (a) presents the daily sectoral NO_x emissions from the source sectors of power, industry, transport, and residential. Fig (b) presents the daily CO₂ to NO_x emission ratio in 2019.

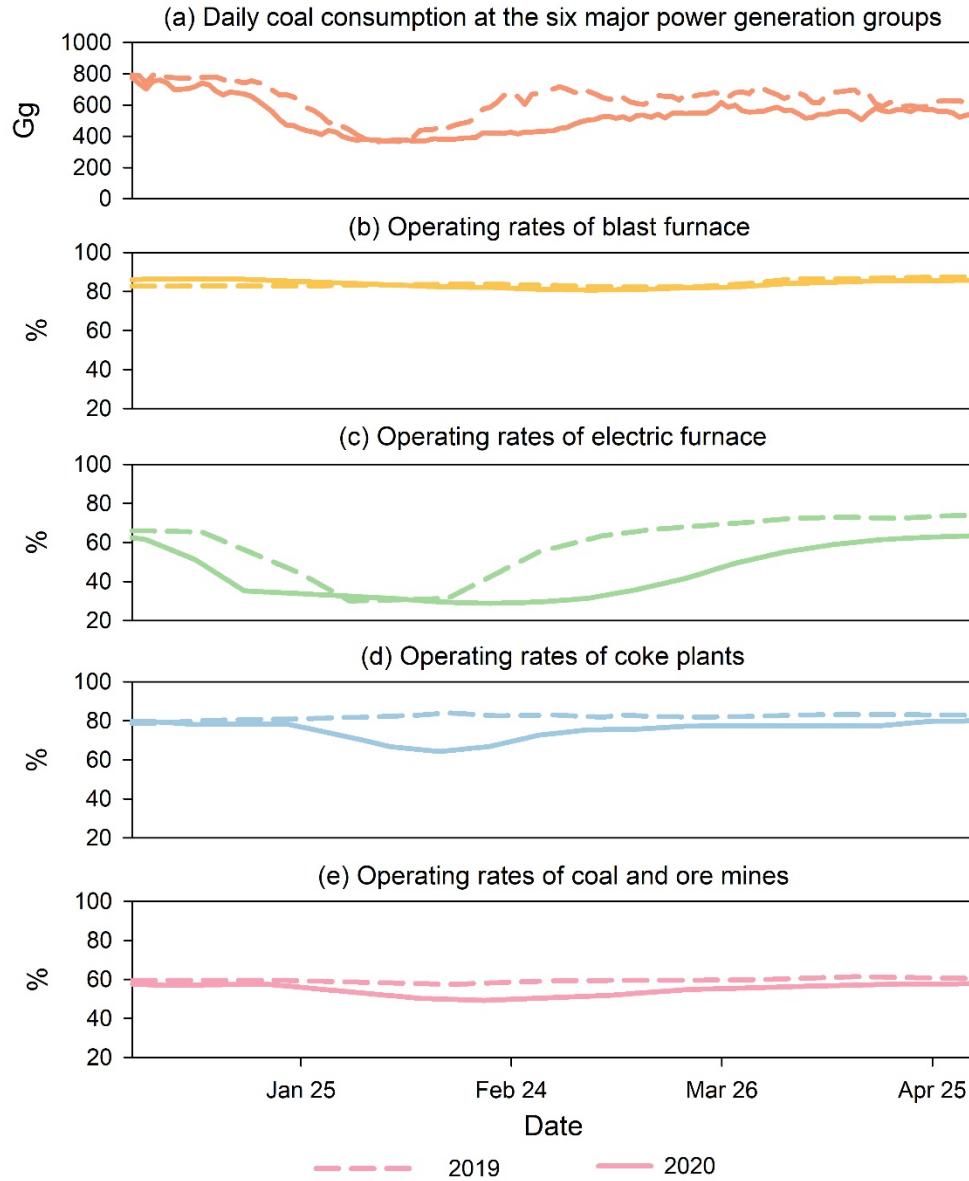


Fig. S3. Daily activities of power and industry sectors from January to April in 2019 and 2020. Fig. (a) presents daily coal consumption at the six major power generation groups. Fig. (b)-(e) present the operating rates of blast furnace, electric furnace, coke plants, and coal and ore mines.

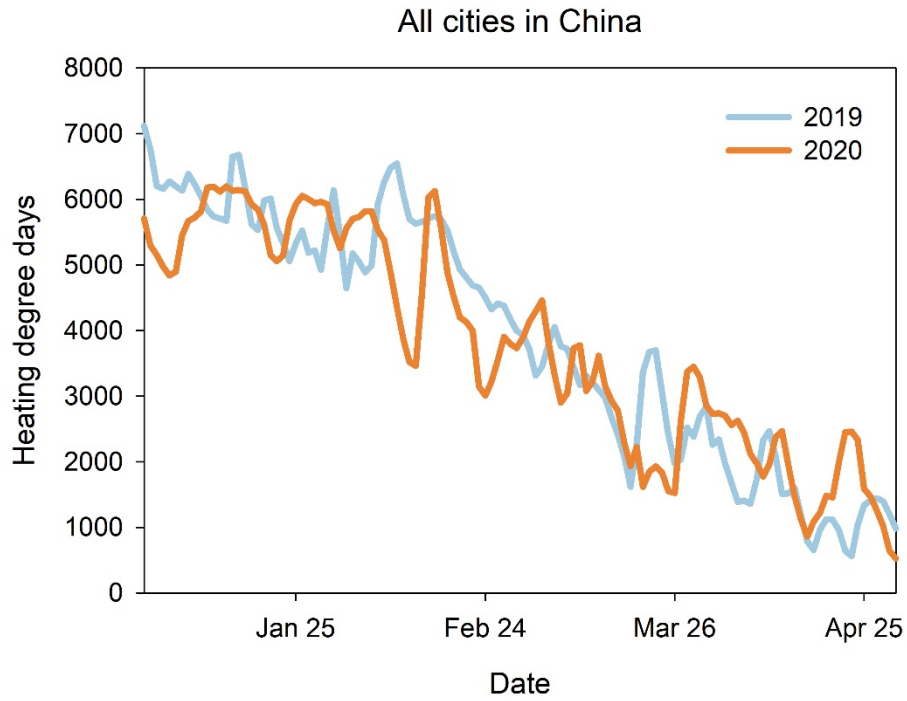


Fig. S4. Population-weighted heating degree days of all cities in China. The data are estimated for each city at a daily scale with the reference temperature of 18°C. The curves represent the sum of the population-weighted heating degree days of all the cities in 2019 and 2020.

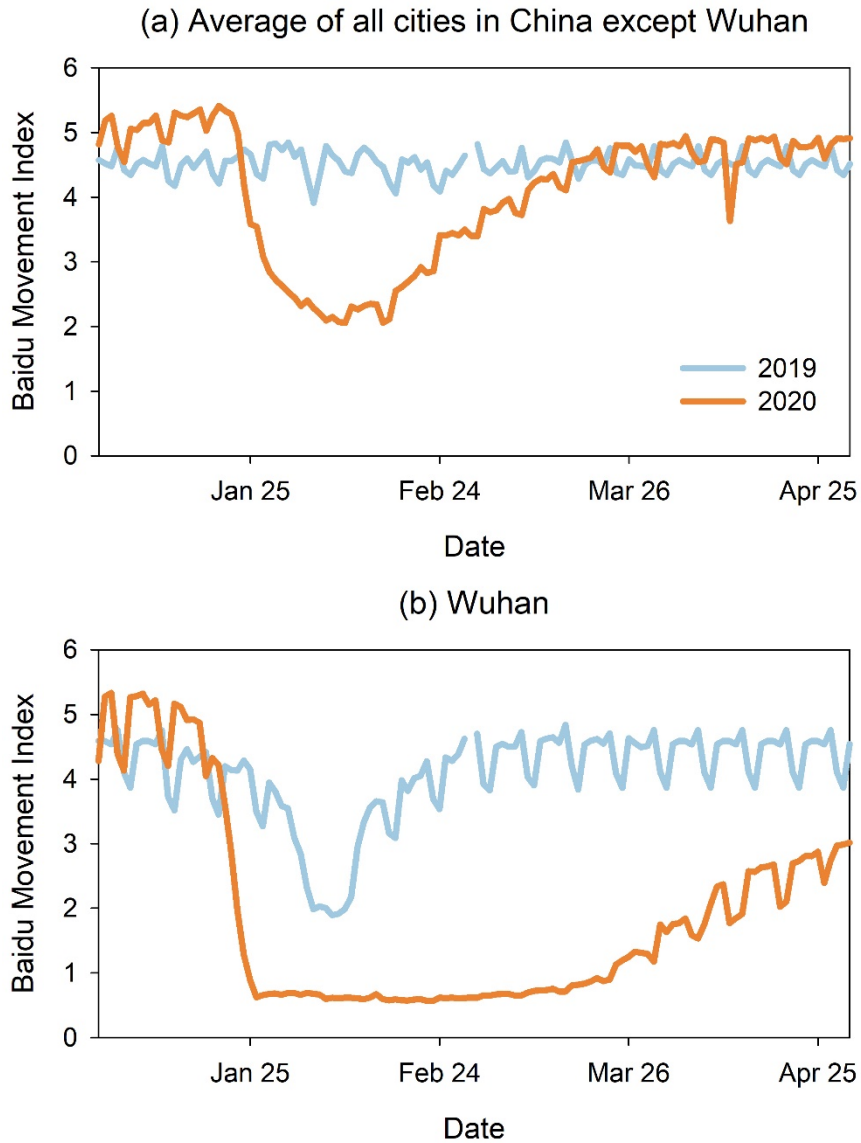


Fig. S5. Daily population migration data within city from the Baidu location-based services. (a) the average of all of China's cities except Wuhan, and (b) Wuhan. The Baidu migration data are derived from the website <https://qianxi.baidu.com/2020/>.

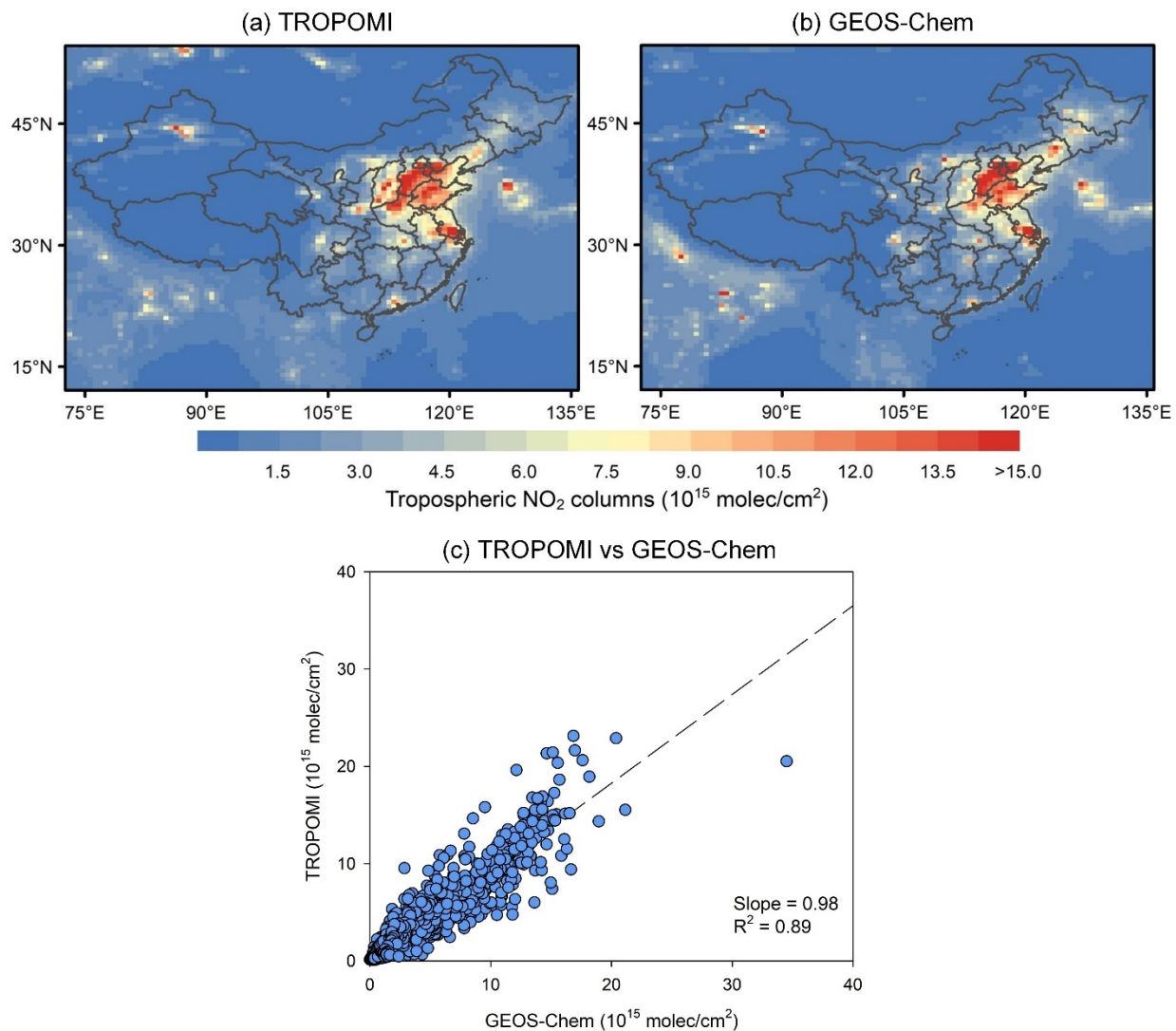


Fig. S6. Evaluation of the GEOS-Chem baseline scenario with NO₂ TVCDs from TROPOMI. Spatial distribution of coincidentally sampled four-month averaged NO₂ TVCDs from (a) TROPOMI and (b) GEOS-Chem. (c) Scatter plot between model simulations and TROPOMI retrievals, with regression R² and slope displaced in the panel.

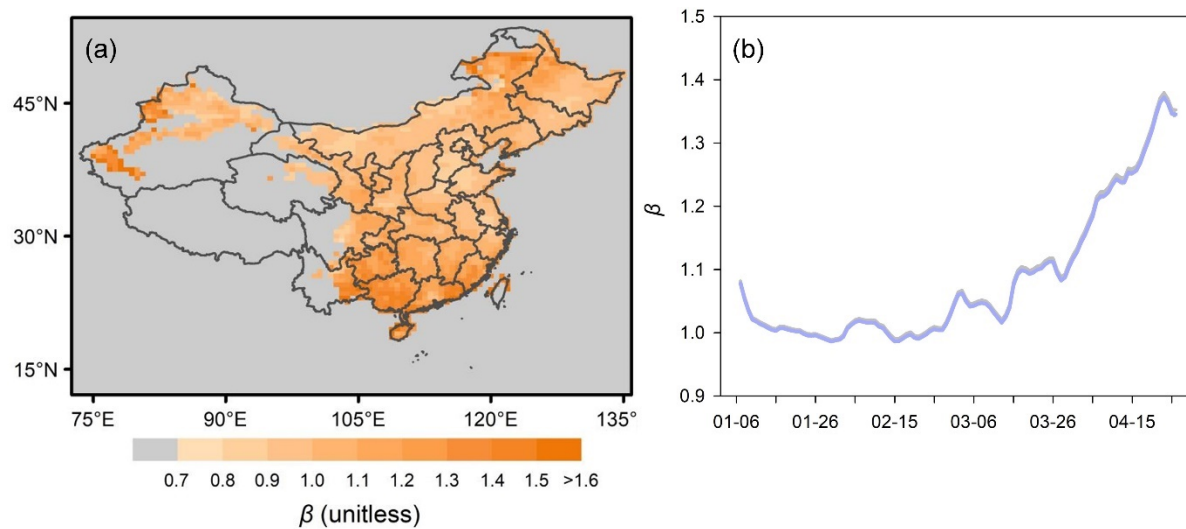


Fig. S7. (a) Spatial distribution and (b) daily variations of beta values over China estimated from GEOS-Chem model. The daily variations are also 10-day moving average.

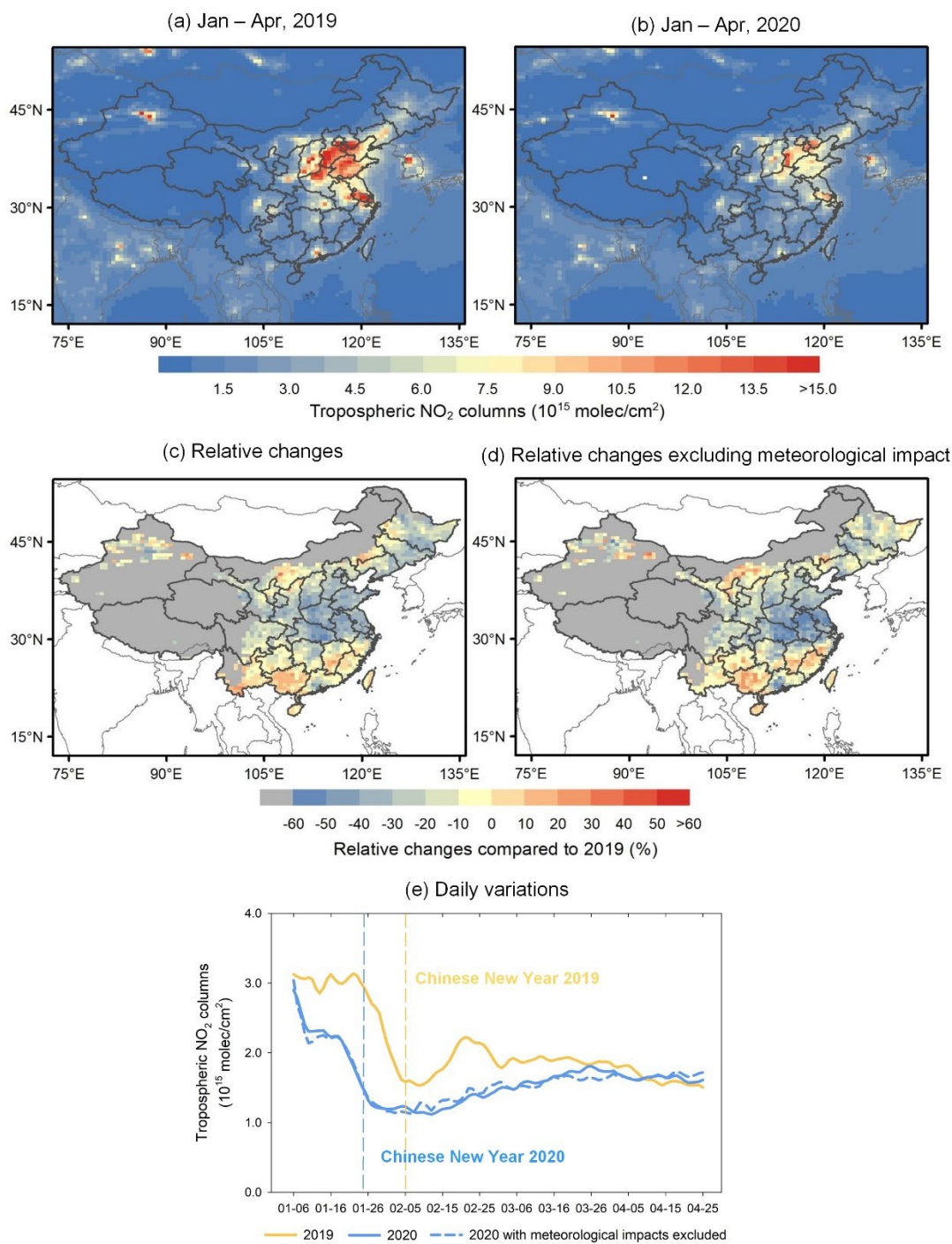


Fig. S8. Tropospheric NO₂ column densities from TROPOMI over China in 2019 and 2020. Average NO₂ TVCDs from Jan to Apr in (a) 2019 and (b) 2020. (c) Total relative differences between 2019 and 2020 and (d) Relative change between 2019 and 2020 excluding the influence from meteorological variations. (e) Daily variations in the 10-day moving average of NO₂ TVCDs from TROPOMI over China.

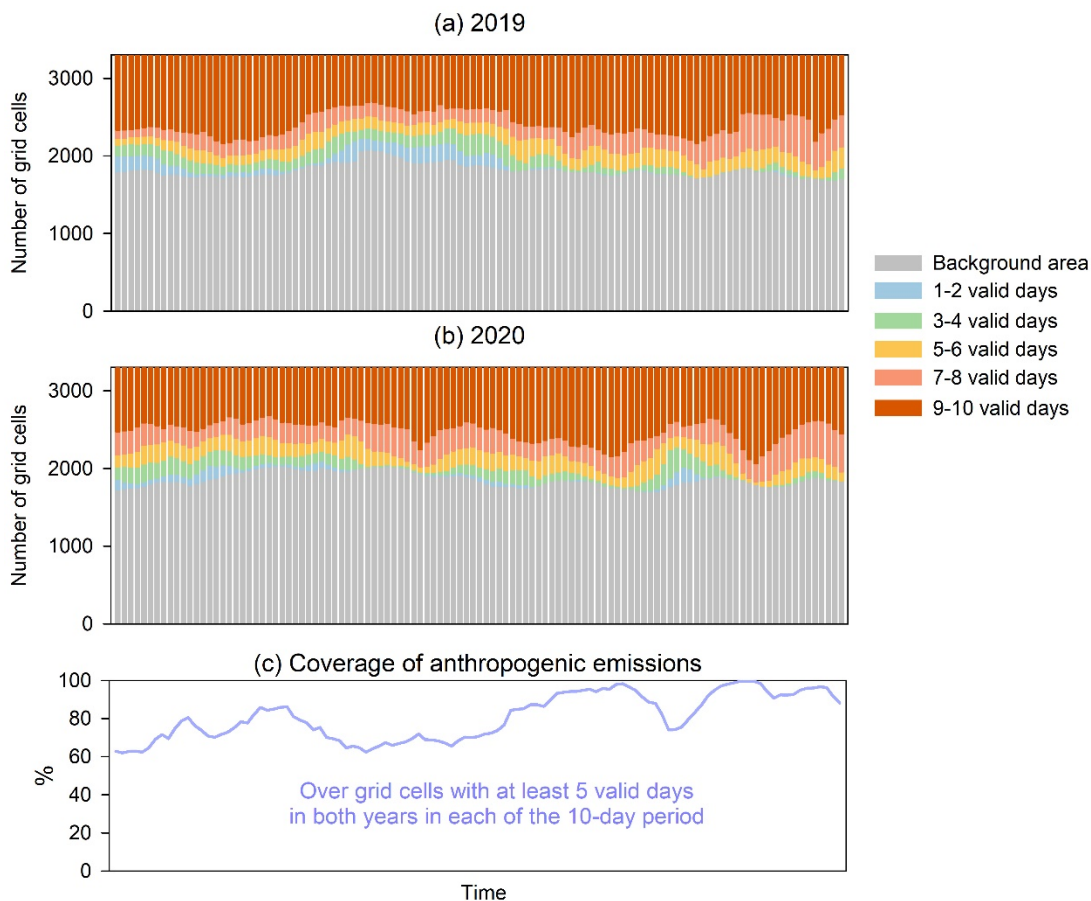


Fig. S9. Sample size in the 10-day moving average calculation of satellite NO₂ TVCDs and representation of national NO_x emissions. Number of grid cells that are filtered out (i.e., $<1 \times 10^{15}$ molec/cm²) (grey) or have a certain number of valid days in (a) 2019 and (b) 2020. (c) The coverage of anthropogenic emissions on each day over grid cells with at least 5 valid days in both years in the 10-day moving average calculation.

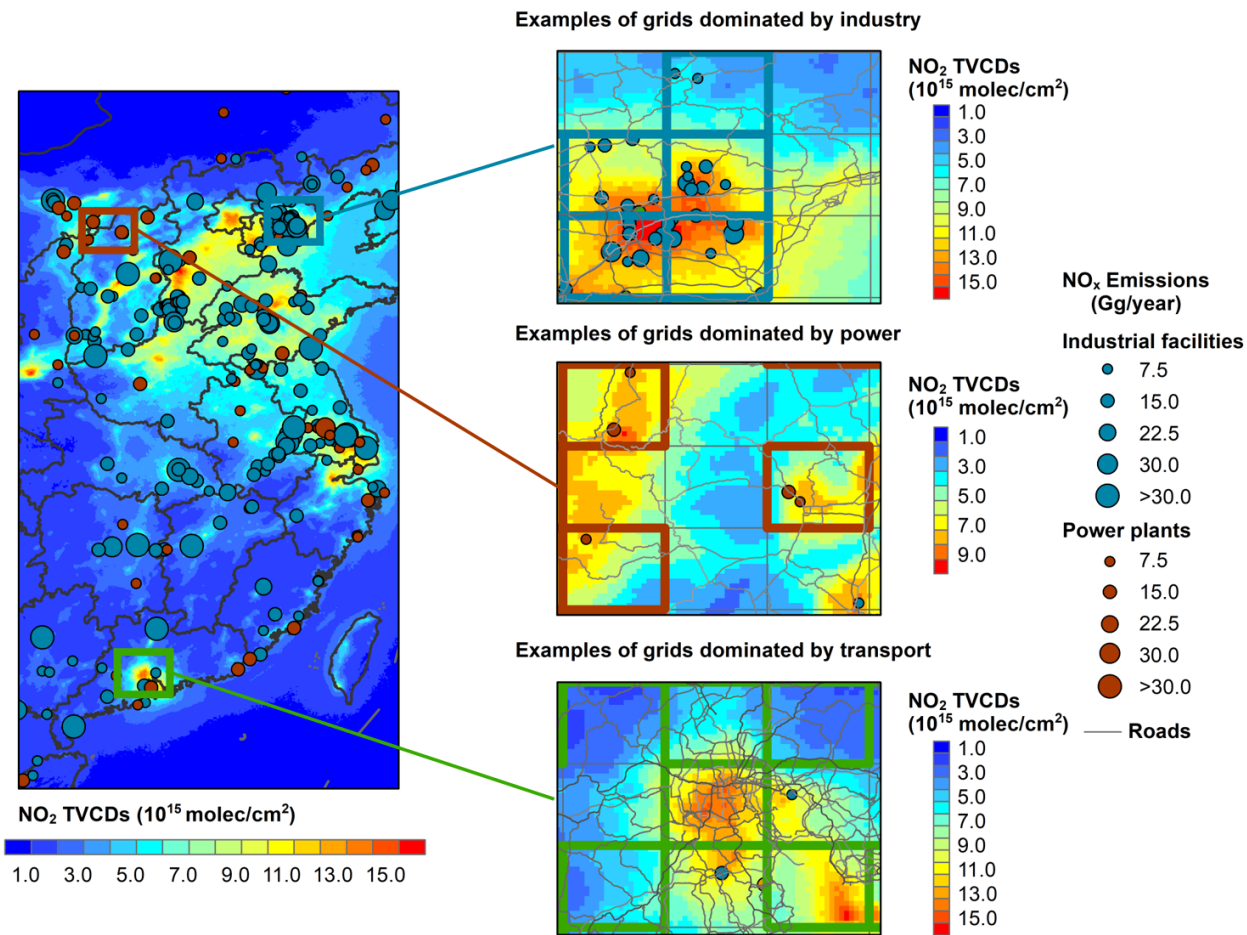


Fig. S10. TROPOMI NO₂ columns over the GEOS-Chem grid cells dominated by industry, power, and transport sectors. The dots represent the point sources of NO_x emissions from industrial facilities and power plants. We also show road networks in the figures.

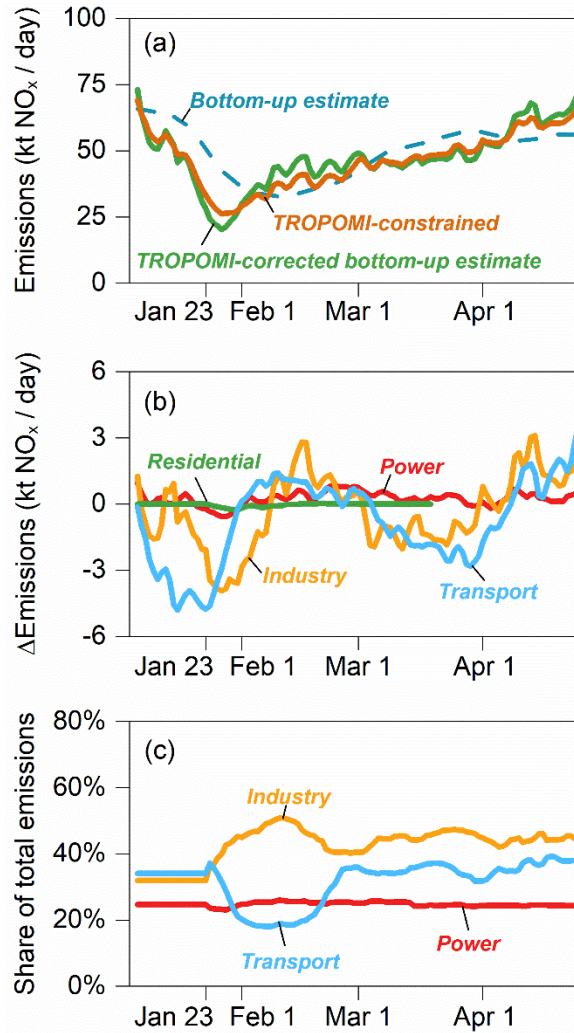


Fig. S11. Comparison of bottom-up NO_x emissions with the TROPOMI inversions in 2020. (a) Comparison of bottom-up estimated NO_x emissions, TROPOMI-constrained NO_x emissions, and TROPOMI-corrected bottom-up estimate. (b) The difference in NO_x emissions between the TROPOMI-constrained estimates and the bottom-up estimates over the grid cells dominated by each source sector. (c) The share of emissions from the grid cells dominated by one source sector in the total emissions of this sector. All of the data are shown in the 10-day moving average.

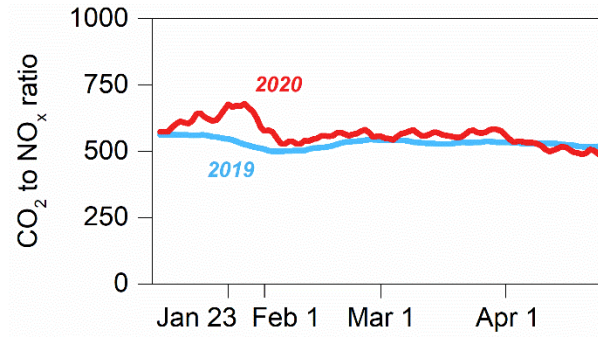


Fig. S12. Comparison of 10-day mean CO₂ to NO_x emission ratio between 2019 and 2020.
The unit is g CO₂/g NO₂.

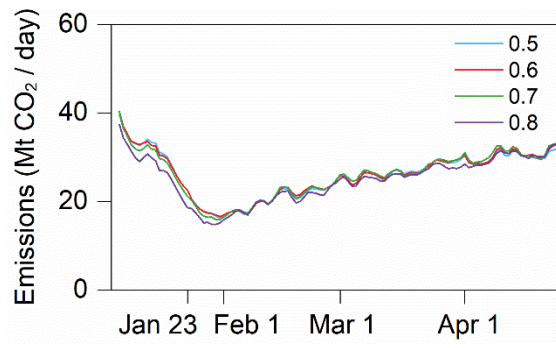


Fig. S13. The comparison of CO₂ inversion emissions derived from different threshold values for the definition of the dominant emission source sector.

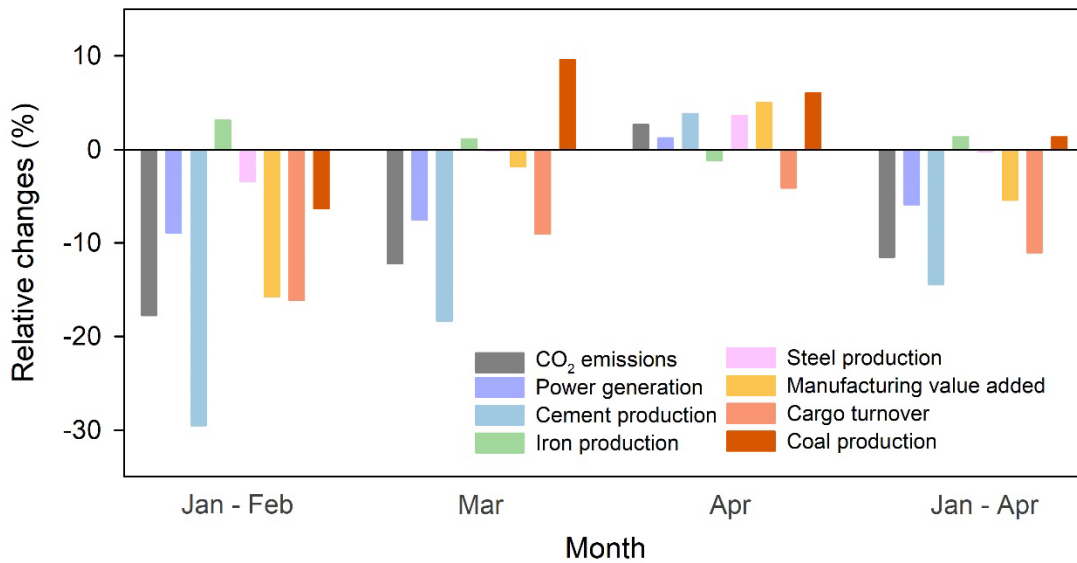


Fig. S14. Relative changes in CO₂ emissions and industrial activities from 2019 to 2020. The CO₂ emissions are derived from the TROPOMI-constrained emissions in this study. The other data that represent the industrial activities in China are all derived from the National Bureau of Statistics (<http://www.stats.gov.cn/>). The comparison between emissions and industrial indicators is made on monthly time scales for Jan-Feb, Mar, Apr, and Jan-Apr.

Table S1. Daily average emissions of NO_x and CO₂ for each month in 2019 and 2020

Month	NO_x (Mt NO₂)			CO₂ (Mt CO₂)		
	2019	2020	Diff	2019	2020	Diff
January	1.6	1.1	-26.9%	854.3	704.3	-17.6%
February	1.4	1.1	-23.3%	728.3	598.0	-17.9%
March	1.8	1.5	-17.0%	948.8	833.6	-12.1%
April	1.4	1.5	5.1%	736.2	755.8	2.7%
January to April	6.1	5.2	-15.9%	3267.6	2891.6	-11.5%

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