



## Supplementary Materials for

### **Unexpected air pollution with marked emission reductions during the COVID-19 outbreak in China**

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## Materials and Methods

### Satellite product

The TROPOMI instrument onboard the Copernicus Sentinel-5P satellite provides daily global coverage of tropospheric column density of NO<sub>2</sub> with a spatial resolution  $3.5 \times 7 \text{ km}^2$  ( $3.5 \times 5.5 \text{ km}^2$  starting from August 6, 2019) with overpass time around 13:30. The TROPOMI NO<sub>2</sub> processing system is based on the algorithm developments for the DOMINO-2 product and for the EU QA4ECV NO<sub>2</sub> reprocessed dataset for OMI, and has been adapted for TROPOMI. In this study, the tropospheric NO<sub>2</sub> column densities were taken from S-5P/TROPOMI Level 2 offline products. The measurements with a Quality Assurance (QA) value less than 0.5 were omitted, which removed the measurements with processing errors, anomalously high signals, and sun glints. We re-gridded the Level 2 product on a  $0.03^\circ \times 0.03^\circ$  spatial grid which corresponds to 3.3 km in latitude.

The MODIS Level 3 AOD datasets are from Aqua (MYD08\_D3) and Terra (MOD08\_D3) with  $1^\circ \times 1^\circ$  spatial resolution. We generate the Level 3 daily map by using the average value of Aqua and Terra measurements. Data can be downloaded from [ftp://ladsweb.nascom.nasa.gov/allData/51/MYD08\\_D3](ftp://ladsweb.nascom.nasa.gov/allData/51/MYD08_D3) and [ftp://ladsweb.nascom.nasa.gov/allData/51/MOD08\\_D3](ftp://ladsweb.nascom.nasa.gov/allData/51/MOD08_D3), respectively.

### Ground station observations

Hourly air quality data have been obtained from the national urban air quality real-time platform released by the China National Environmental Monitoring Station (website: <http://106.37.208.233:20035>). Information from the national air quality monitoring stations used in this study over four major cities is shown in Table S1. The method of measuring air quality follows the national standard of GB 3095-2012. Measurement of PM<sub>2.5</sub> was carried out by the  $\beta$  Ray absorption method and micro-oscillation balance method; measurement of NO<sub>2</sub> is by the Chemiluminescence method; measurement of SO<sub>2</sub> and O<sub>3</sub> is by the UV fluorescence method; measurement of CO is by the non-dispersive infrared absorption method and gas filter correlation infrared absorption method. Monitoring stations usually avoid tall buildings, trees and other potential obstacles that would impede air circulation. The surroundings of air monitoring site have been guaranteed with stable electricity supplies and device maintenance service. Sampling ports are 3 to 15 meters above the ground. The distance between each sampling port exceeds 1 m. Temperature inside the monitoring stations is maintained between 15 and 35 °C, relative humidity  $\leq 85\%$ , and atmospheric pressure between 80 and 106 kPa.

### Reanalysis data

The boundary layer height, precipitation, relative humidity at 1000 hPa and the wind vector  $w = (u,v)$  at 10 meters above the ground were taken from the ERA5 reanalysis data with  $0.25^\circ \times 0.25^\circ$  spatial resolution. ERA5 combines historical observations into global estimates using advanced modelling and data assimilation systems.

### WRF-Chem model simulations

The model used in this study is based on a specific version of the WRF-Chem model (23) with modification by (24–27). The specific WRF-Chem model includes a flexible gas phase chemical module with consideration of different chemical mechanisms and the CMAQ aerosol module (AERO5) developed by US EPA (28). The organic aerosols (OA) are simulated using the volatility basis-set (VBS) modeling method, with the secondary OA (SOA) contributions from glyoxal and methylglyoxal. ISORROPIA (Version 1.7) is used to predict the inorganic aerosols, calculating the composition and phase state of an ammonium-sulfate-nitrate-water inorganic aerosol in thermodynamic equilibrium with gas phase precursors (29). Three major types of heterogeneous aerosol chemistry are considered: the heterogeneous hydrolysis of  $N_2O_5$  on the surface of deliquescent aerosols to form nitrate, the heterogeneous reaction of  $SO_2$  involving aerosol water to form sulfate, and the heterogeneous reaction of glyoxal and methylglyoxal to form SOA.

The anthropogenic emission inventory is developed by (30) with the base year of 2020, including industry, transportation, power plant, residential and agriculture sources. The Model of Emissions of Gases and Aerosols from Nature (MEGAN) is used to calculate the biogenic emissions online (31). Biomass burning emissions are potentially important for Asian haze (32–35). Fire counts from satellites can partly reflect BB in agricultural activities and wildfires. In the fig. S9, we analyzed the MODIS fire counts during the 2020 city-lockdown period and found very few fire activities in the Beijing-Tianjin-Hebei area. Moreover, the fires were more frequent in the same period of 2019 than that in 2020 over the whole nation. Therefore, we conclude that BB from agriculture and wildfires did not contribute significantly to the haze formation during the 2020-CLD, and we did not consider them in our WRF-Chem simulations. A haze episode from 21 January to 16 February 2020 in the North China Plain is simulated using the WRF-Chem model, and detailed model configuration can be found in Table S2. A series of model sensitivity experiments are conducted, and the experiment descriptions are provided in Table S3.

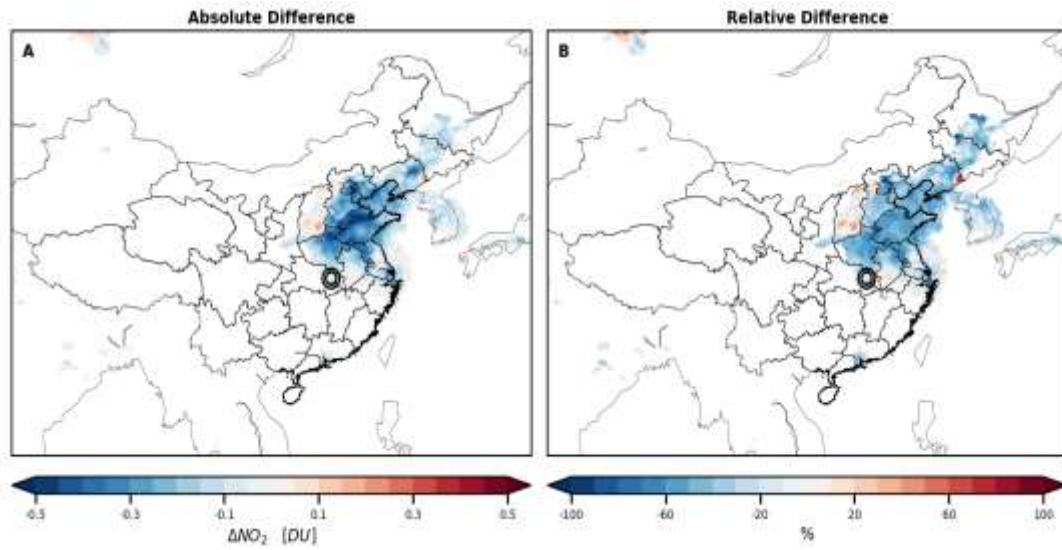
The mean bias (MB), root mean square error (RMSE) and the index of agreement (IOA) are used to evaluate the model performance in simulating air pollutants.

$$MB = \frac{1}{N} \sum_{i=1}^N (P_i - O_i)$$

$$RMSE = \left[ \frac{1}{N} \sum_{i=1}^N (P_i - O_i)^2 \right]^{\frac{1}{2}}$$

$$IOA = 1 - \frac{\sum_{i=1}^N (P_i - O_i)^2}{\sum_{i=1}^N (|P_i - \bar{O}| + |O_i - \bar{O}|)^2}$$

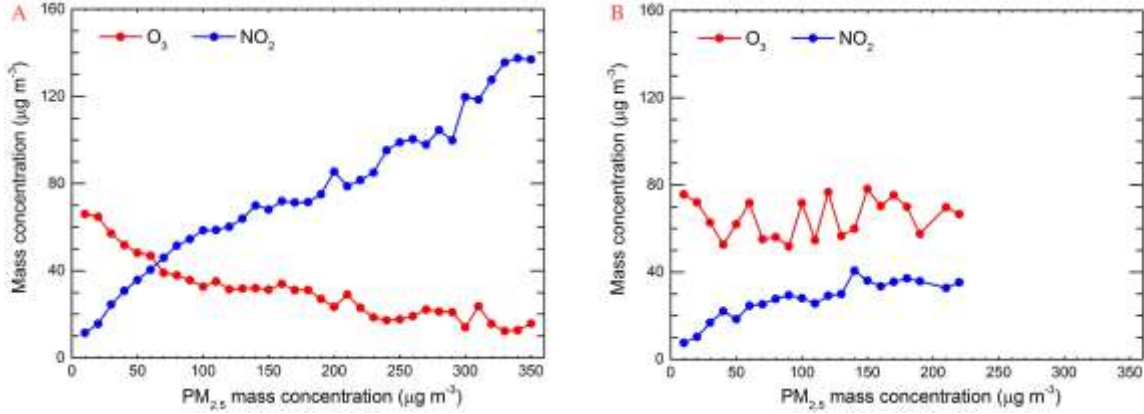
Where  $P_i$  and  $O_i$  are the simulated and observed variables, respectively.  $N$  is the total number of the simulations for comparisons, and  $\bar{O}$  donates the average of the observations.



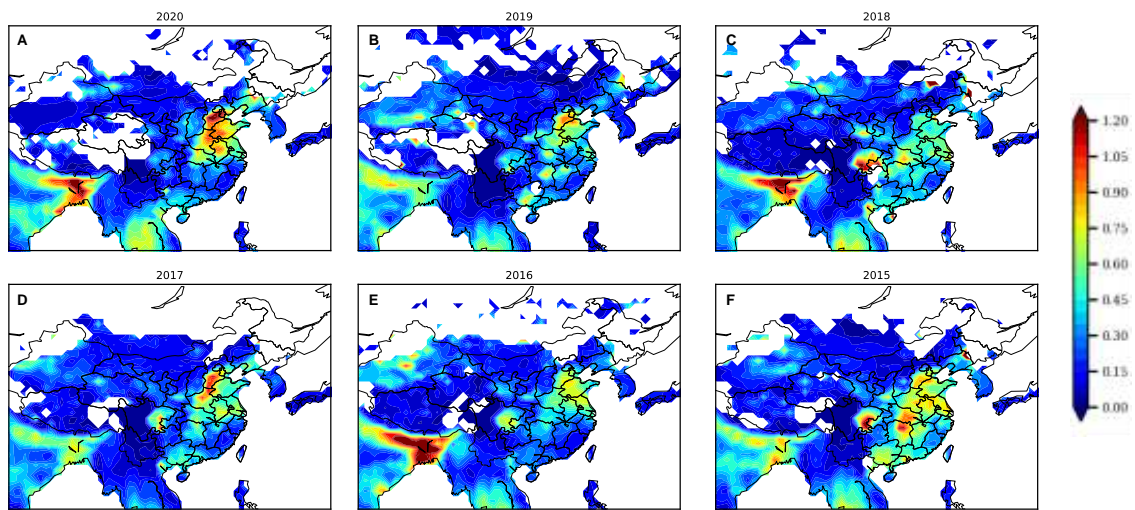
**Fig. S1.** Comparison of column integrated NO<sub>2</sub> between three weeks during the 2020 lockdown and 2015-2019 climatology using NASA Aura OMI products. Only regions with background NO<sub>2</sub> larger than 0.2 DU are analyzed.



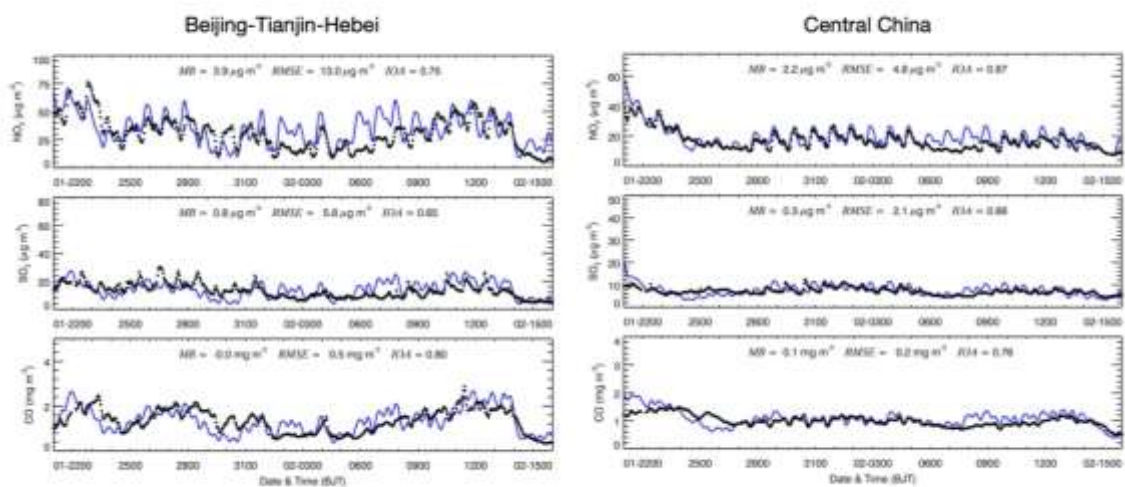
**Fig. S2.** Spatial distribution of the 1515 state monitoring stations.



**Fig. S3.** Daytime variation of ozone and  $\text{NO}_2$  (10:00 to 16:00 LST) as a function of  $\text{PM}_{2.5}$  in the winter from 2015 to 2019 (A) and during 2020-CLD (B). During wintertime in north China, the weak insolation slows the atmospheric photochemistry processes. Therefore, very high  $\text{NO}_x$  emissions in the region cause remarkable ozone titration even during daytime, as shown in A, particularly with increasing  $\text{PM}_{2.5}$  which further attenuates the incoming solar radiation. However, during 2020-CLD, significant decrease in  $\text{NO}_x$  emissions alleviates the ozone titration, and during haze episodes, the ozone level is much higher than that during the winter from 2015 to 2019. The ozone concentration fluctuates at around  $65 \mu\text{g m}^{-3}$  with  $\text{PM}_{2.5}$  exceeding  $35 \mu\text{g m}^{-3}$ , caused by complicated nonlinear ozone chemistry.

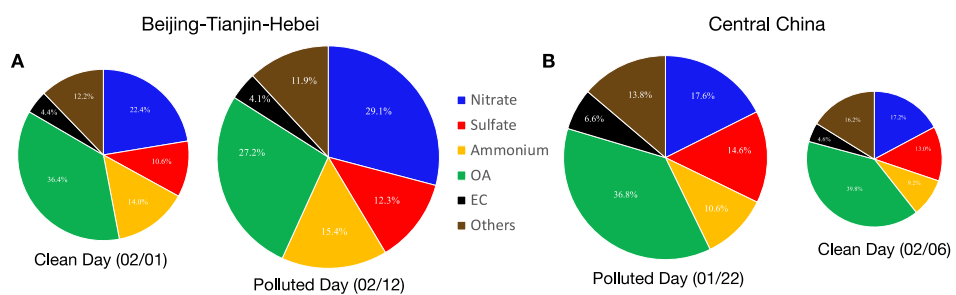


**Fig. S4.** MODIS L3 AOD from Terra and Aqua during the same three-week period with 2020-CLD from 2015 to 2020.

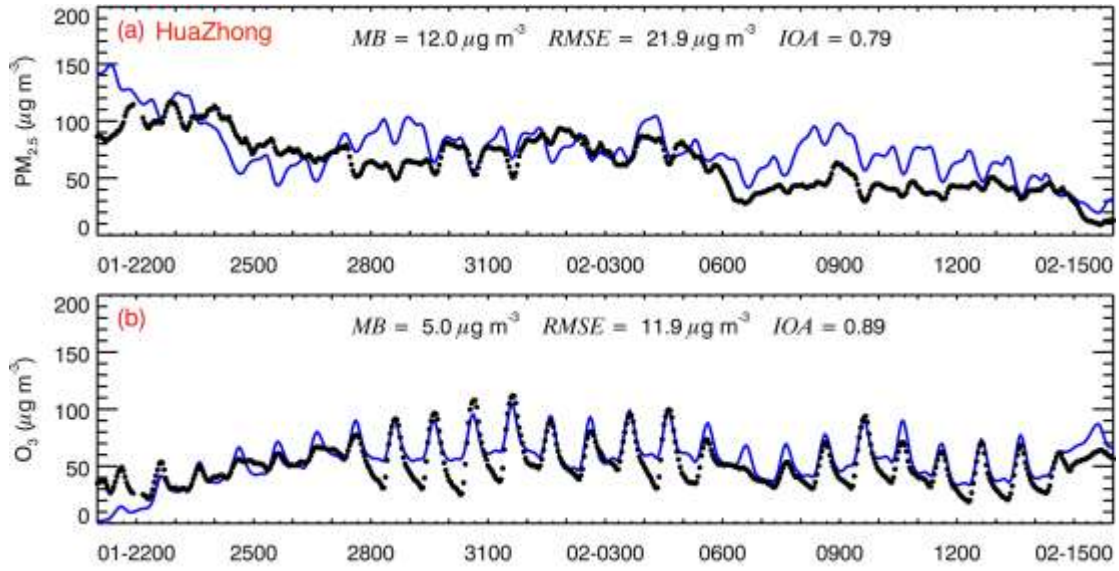


**Fig. S5.** Comparison of observed (black dots) and simulated (blue line) diurnal profiles of near surface hourly NO<sub>2</sub>, SO<sub>2</sub>, and CO averaged over all ambient monitoring stations in Beijing-Tianjin-Hebei (left panels) and Central China (Henan-Hubei-Hunan, right panels) from 21 January to 16 February 2020.

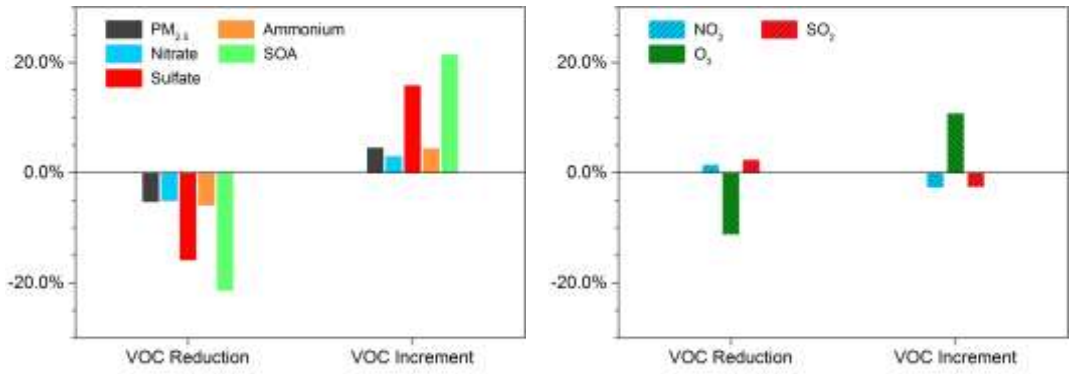




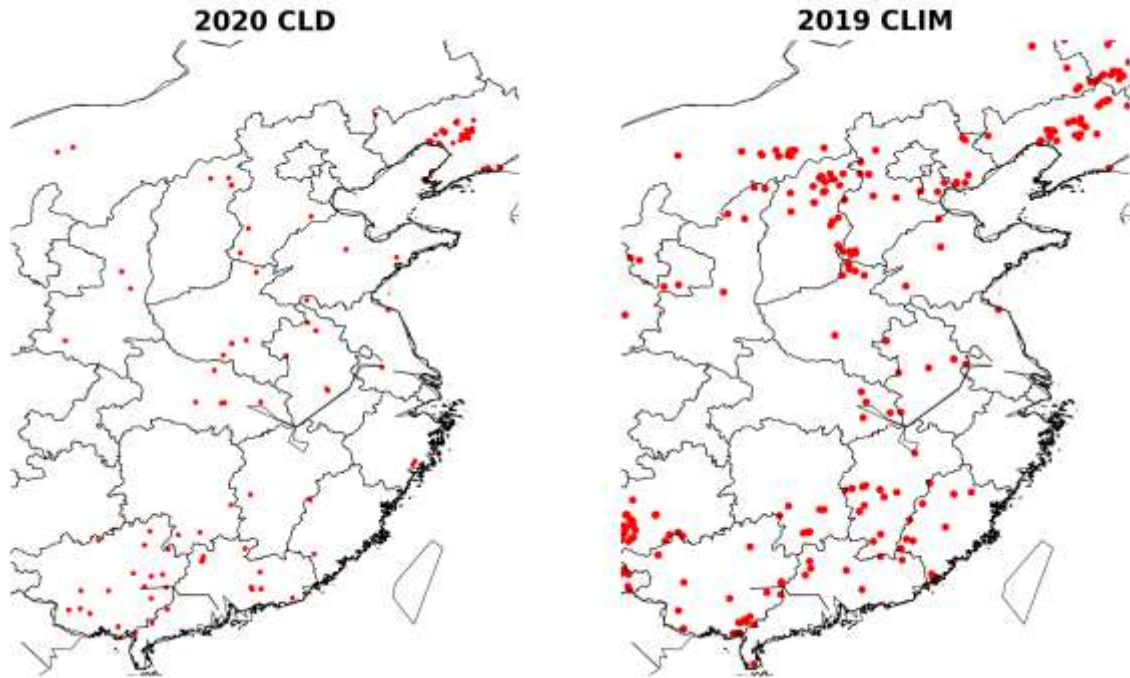
**Fig. S6.** WRF-Chem simulated surface aerosol species over two characteristic regions: Beijing-Tianjin-Hebei in northern China and Central China. OA and EC are short for organic aerosol and elementary carbon, respectively.



**Fig. S7.** WRF-Chem simulated time evolutions of PM<sub>2.5</sub> and ozone in Central China. Black dots are observations, and blues lines are model simulations.



**Fig. S8.** Simulated sensitivity of aerosol species and precursor gases to the VOC changes by 30%.



**Fig. S9.** Accumulated fire counts based on MODIS Active Fire Products over three-week periods during 2020-CLD and 2019-CLIM. Fire counts can be used to indicate the biomass burning in the agricultural activities. Dots in the plots are with confidence level larger than 80 in the MODIS product.

**Table S1.** National Air Quality Monitoring Stations

Station ID	Station Name	City	Longitude	Latitude	Station ID	Station Name	City	Longitude	Latitude
1001 A	Wan Shou Xi Gong	Beijing	116.366	39.867	1325 A	Dong Hu Li Yuan	Wuhan	114.367	30.572
1002 A	Ding Ling	Beijing	116.17	40.287	1326 A	Han Yang Yue Hu	Wuhan	114.251	30.551
1003 A	Dong Si	Beijing	116.434	39.952	1327 A	Han Kou Hua Qiao	Wuhan	114.284	30.62
1004 A	Temple of Heaven	Beijing	116.434	39.875	1328 A	Wu Chang Zi Yang	Wuhan	114.301	30.549
1005 A	Agriculture Exhibition Center	Beijing	116.473	39.972	1329 A	Qing Shan Gang Hua	Wuhan	114.427	30.61
1006 A	Guan Yuan	Beijing	116.361	39.943	1330 A	Tun Kou New District	Wuhan	114.153	30.475
1007 A	Wan Liu Haidian District	Beijing	116.315	39.993	1331 A	Han Kou Jiang Tan	Wuhan	114.301	30.595
1008 A	Xincheng Shunyi	Beijing	116.72	40.144	1332 A	Dong Hu Gao Xin	Wuhan	114.389	30.482
1009 A	Huai Rou Town	Beijing	116.644	40.394	1333 A	Wu Jia Shan	Wuhan	114.213	30.641
1010 A	Chang Ping Town	Beijing	116.23	40.195	1334 A	Chen Hu Qi Hao	Wuhan	113.853	30.3
1011 A	Olympic Center	Beijing	116.407	40.003	1345 A	Guang Ya Middle School	Guangzhou	113.235	23.142
1012 A	Ancient city	Beijing	116.225	39.928	1346 A	Guangzhou No.5 High School	Guangzhou	113.261	23.105
1141 A	Pu Tuo	Shanghai	121.4	31.238	1348 A	Guang Dong Business College	Guangzhou	113.348	23.092
1142 A	Shi Wu Chang	Shanghai	121.478	31.204	1349 A	Guangzhou No.86 High School	Guangzhou	113.433	23.105
1143 A	Hong Kou	Shanghai	121.467	31.301	1350 A	Fan Yu Middle School	Guangzhou	113.352	22.948
1144 A	Shanghai Normal University in Xuhui	Shanghai	121.412	31.165	1351 A	Hua Du Normal University	Guangzhou	113.215	23.392

1145 A	Yang Pu Si Piao	Shangha i	121.536	31.266	1352 A	Guangzho u Monitorin g Station	Guangzho u	113.26	23.133
1146 A	Qing Pu Dian Shan Lake	Shangha i	120.978	31.094	1353 A	Jiu Long Town Zhen Long	Guangzho u	113.568	23.278
1147 A	Jing An Monitorin g Station	Shangha i	121.425	31.226	1354 A	Lu Hu	Guangzho u	113.281	23.157
1148 A	Pu Dong Chuan Sha	Shangha i	121.703	31.191	1355 A	Mao Feng Shan Forestry Park	Guangzho u	113.589	23.554
1149 A	Pu Dong New District Monitorin g Station	Shangha i	121.533	31.228	2846 A	Ti Yu Xi	Guangzho u	113.321	23.132
1150 A	Pu Dong Zhang Jiang	Shangha i	121.577	31.207					

**Table S2.** WRF-Chem model configurations.

Region	East Asia
Simulation period	2020-01-21 to 2020-02-16
Domain size	400 × 400
Domain center	35.0°N, 114.0°E
Horizontal resolution	12km × 12km
Vertical resolution	35 vertical levels with a stretched vertical grid with spacing ranging from 30m near the surface, to 500m at 2.5km and 1km above 14km
Microphysics scheme	WSM 6-class graupel scheme (36)
Boundary layer scheme	MYJ TKE scheme (37)
Surface layer scheme	MYJ surface scheme (37)
Land-surface scheme	Unified Noah land-surface model (38)
Long-wave radiation scheme	Goddard longwave scheme (39)
Short-wave radiation scheme	Goddard shortwave scheme (40)
Meteorological boundary and initial conditions	NCEP 1°×1° reanalysis data
Chemical initial and boundary conditions	MOZART 6-hour output (41)
Anthropogenic emission inventory	SAPRC-99 chemical mechanism emissions (30)
Biogenic emission inventory	MEGAN model developed by (31)
Model spin-up time	24 hours

**Table S3.** Model sensitivity experiment description.

<i>Experiments</i>	<i>Configuration</i>	<i>Purpose</i>
Baseline	Described in Table S2.	To reproduce observed pollution changes.
Clim_Met	Using Climatological meteorological initial and boundary conditions averaged over the same time periods during 2015 - 2019.	To assess the meteorological influence on pollution changes.
NOx_80	According to the satellite observations, increasing NOx emissions in all sectors by 80% from the baseline simulation to reflect the non-COVID19 scenario.	To assess effect of NOx reduction.
Heteoro_Chem	Turning off all heterogeneous chemistry processes in our modified version of WRF-Chem.	To assess the contribution of heterogeneous chemistry to the haze formation.
VOC	Increasing/Decreasing VOC emissions by 30%.	To assess the sensitivity of VOC emissions to the haze formation.



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