Dear Reviewer,

We appreciate your comments and suggestions, which have helped us improve our manuscript further. We have made the necessary changes to the manuscript, which can be found in the attached file (Track Changes). The following is a response to your comments and suggestions. Corresponding changes in the revised manuscript are also made available below, if applicable, at the appropriate places.

Sincerely,

On behalf of all co-authors, Vigneshkumar Balamurugan

The manuscript entitled "Secondary PM decreases significantly less than NO2 emission reductions during COVID lockdown in Germany" by Vigneshkumar Balamurugan et al. explored the drivers of slight decrease of PM2.5 compared to NO2 emission during COVID-19 lockdown in Germany. The manuscript provides valuable information for understanding PM pollution under rigorous emission reduction measures and efficiently directing PM mitigation in the future. It is recommended that this manuscript be reconsidered for publication after major revisions.

Thank you so much for reading and reviewing our manuscript! We carefully reviewed and considered your comments/suggestions, and made improvements in the revised manuscript.

General comments:

Line 54:" The composition of PM thus varies greatly depending on time and location; for example, in urban areas nitrate and organic aerosol often dominate in winter time". More cases should be given to support this sentence.

We cited additional studies to support this statement.

Lines 55-58: The composition of PM thus varies greatly depending on time and location; for example, in urban areas nitrate and organic aerosol often dominate in winter time (Cesari et al., 2018; JudaRezler et al., 2020; Samek et al., 2020; Salameh et al., 2015; Womack et al., 2019; Zhai et al., 2021).

Line 133:" The fractional change in meteorology accounted for pollutant concentration between 2020 and 2019, i.e., pollutant concentration changes between 2020 and 2019 due to emission changes only" This definition is misleading. According to your definition of $\Delta PM2.5$ (obs) and $\Delta PM2.5$ (GC), the

ΔPM2.5(obs,emi) should be the change of PM2.5 caused only by emission. If so, relative descriptions in the whole paper should be revised correspondingly.

 $\Delta PM_{2.5(obs,emi)}$ = Absolute concentration changes (µg m⁻³) after accounting for meteorology (caused only by emission) between 2020 and 2019.

 $f \, \text{PM}_{2.5 \text{(obs, emi)}} = \text{Fractional concentration changes (\%)}$ after accounting for meteorology (caused only by emission) between 2020 and 2019.

We hope this clarifies your comment. We also made minor changes to the sentence and equation to make it clearer to the reader.

| Lines 139-141: | The fractional change in meteorology accounted for pollutant concentration between 2020 and 2019, i.e., fractional change (%) in pollutant concentration between 2020 and 2019 due to emission changes only, is calculated as, |
|-------------------|--|
| | $fPM_{2.5(obs,emi)} = (\Delta PM_{2.5(obs,emi)} / PM_{2.5(obs,2019)}) * 100$ |

Line 170: We also compared the 2019 GC and 2019 observed in-situ PM2.5 concentrations and found that the GC and observed in-situ PM2.5 concentrations were in good agreement (R > 0.5 for all metropolitan areas, except Leipzig which has a R value of 0.39) (e.g.,Fig. 6 (c), for Cologne metropolitan area)." The performance of the model is the base of further analysis. Hence, more details of the statistical evaluation of the model performance for each site should be given. In addition, the agreement R is above 0.5 for most areas and is 0.39 for Leipzig. Personally, I think the R is not good enough.

In the revised manuscript (Table A1), we now included the performance of the GC model for each metropolitan area. The R value in all cases is between 0.3 and 0.7, indicating moderate correlation (https://link.springer.com/article/10.1057/jt.2009.5).

Table A1. The statistical evaluation (R, RMSE and mean bias) of the GC model performance (PM_{2.5}) for the 2019 study period (January 1 to May 31).

| Metropolitan area | Correlation coefficient (R) | RMSE (µg m ⁻³) | Mean bias (GC – insitu / insitu) (%) | |
|-------------------|-----------------------------|----------------------------|--------------------------------------|--|
| Bremen | 0.6 | 8.7 | -18.9 | |
| Cologne | 0.5 | 11 | 11.7 | |

| Dresden | Dresden 0.56 | | -18.8 |
|------------|----------------|------|-------|
| Dusseldorf | 0.53 | 10.5 | -15.7 |
| Frankfurt | Frankfurt 0.58 | | -37.4 |
| Hamburg | 0.67 | 8 | -12.7 |
| Hanover | 0.59 | 7.9 | -13.1 |
| Leipzig | 0.39 | 8.4 | -28.6 |
| Munich | 0.5 | 8.5 | -18.6 |
| Stuttgart | 0.53 | 8.6 | -16.1 |

Line 273:" The increase in OH radicals results from German metropolitan areas being in a NOX saturated regime". From BAU to lockdown period, the meteorological condition changed, which could lead to higher temperature and higher solar radiation, and this has the potential to increase OH concentration. Hence, the influence of meteorological between different period in 2020 should be considered.

Section 4.2 shows a comparison of 2020 (lockdown) and 2020 (no lockdown). The meteorology is the same in both cases; the only difference is in the emissions. As a result, the change in OH must be due to chemistry changes caused by changes in NO_X emissions. We hope this clarifies your comment.

Line 281:"However, higher night-time NO3 levels result in higher nighttime HNO3 production from N2O5 hydrolysis, resulting in slightly lower night-time lockdown PM nitrate compared to BAU" According to Figure 4, the change of nighttime HNO3 production from N2O5 hydrolysis is small compared to that during daytime. In addition, both of the production and sink of HNO3 should be considered to explain its influence on PM concentration.

Thanks for pointing this out. We modified the sentence as follow,

| Lines | However, higher night-time NO₃ levels result in relatively unchanged night- |
|----------|--|
| 310-312: | time HNO_3 production from N_2O_5 hydrolysis, resulting in slightly lower night- |
| | time lockdown PM nitrate compared to BAU (Fig. 4 (b,e,f,g)). |

Our GEOS-Chem simulation diagnostics does not allow us to extract the HNO₃ sink information directly. Deposition (dry and wet deposition) is the primary sink of HNO₃ in the troposphere. Since we discuss the difference between 2020 (lockdown) and 2020 (no lockdown), which have the same meteorology with different emission, we don't expect the rate of dry deposition (calculated based on different meteorological variables (http://wiki.seas.harvard.edu/geos-chem/index.php/Dry_deposition) and wet deposition to differ significantly.

Specific comments:

The use of "emission accounted", and "meteorology accounted" makes the discussion part puzzled. The authors are suggested to use more clear phases.

We agree with the reviewer that these words make the discussion a bit confusing sometimes. We discussed this with our other colleagues and chose these words, because they are more clear in terms of methodology than other words. However, we are open to choose if a reviewer is willing to suggest new words.

Figure 1: The part of "Ground-truth measurements" is misleading, it should contain the observations data from 2019 and 2020.

Thanks for the suggestion. We modified the figure 1, as you suggested.

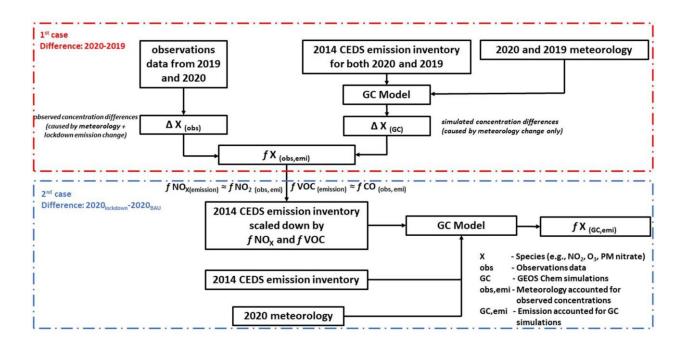


Figure 1. Schematic diagram of our methodology for calculating the meteorology accounted for observed pollutant concentrations changes between 2020 and 2019, and emission accounted for GC pollutant concentrations changes between 2020 lockdown and 2020 BAU scenario.

References:

Cesari, D., De Benedetto, G., Bonasoni, P., Busetto, M., Dinoi, A., Merico, E., Chirizzi, D., Cristofanelli, P., Donateo, A., Grasso, F., et al.: Seasonal variability of PM2. 5 and PM10 composition and sources in an urban background site in Southern Italy, Science of the Total Environment, 612, 202–213, 2018.

Juda-Rezler, K., Reizer, M., Maciejewska, K., Błaszczak, B., and Klejnowski, K.: Characterization of atmospheric PM2. 5 sources at a Central European urban background site, Science of the Total Environment, 713, 136 729, 2020.

Samek, L., Stegowski, Z., Styszko, K., Furman, L., Zimnoch, M., Skiba, A., Kistler, M., Kasper-Giebl, A., Rozanski, K., and Konduracka, E.: Seasonal variations of chemical composition of PM2. 5 fraction in the urban area of Krakow, Poland: PMF source attribution, Air Quality, Atmosphere & Health, 13, 89–96, 2020.

Salameh, D., Detournay, A., Pey, J., Pérez, N., Liguori, F., Saraga, D., Bove, M. C., Brotto, P., Cassola, F., Massabò, D., et al.: PM2. 5 chemical composition in five European Mediterranean cities: a 1-year study, Atmospheric Research, 155, 102–117, 2015.

Womack, C., McDuffie, E., Edwards, P., Bares, R., de Gouw, J., Docherty, K., Dubé, W., Fibiger, D., Franchin, A., Gilman, J., et al.: An odd oxygen framework for wintertime ammonium nitrate aerosol pollution in urban areas: NOx and VOC control as mitigation strategies, Geophysical Research Letters, 46, 4971–4979, 2019.

Zhai, S., Jacob, D. J., Wang, X., Liu, Z., Wen, T., Shah, V., Li, K., Moch, J. M., Bates, K. H., Song, S., et al.: Control of particulate nitrate air pollution in China, Nature Geoscience, pp. 1–7, 2021.

Secondary PM_{2.5} decreases significantly less than NO₂ emission reductions during COVID lockdown in Germany

Vigneshkumar Balamurugan¹, Jia Chen¹, Zhen Qu², Xiao Bi¹, and Frank N. Keutsch^{2,3}

Correspondence: Vigneshkumar Balamurugan (vigneshkumar.balamurugan@tum.de), Jia Chen (jia.chen@tum.de)

Abstract. This study estimates the influence of anthropogenic emission reductions on the concentration of particulate matter with a diameter smaller than 2.5 μm (PM_{2.5}) during the 2020 lockdown period in German metropolitan areas. After accounting for meteorological effects, PM_{2.5} concentrations during the spring 2020 lockdown period were 5 % lower compared to the same time period in 2019. However, during the 2020 pre-lockdown period (winter), meteorology accounted for PM_{2.5} concentrations were 19 % lower than in 2019. Meanwhile, meteorology accounted for NO2 concentrations dropped by 23 % during the 2020 lockdown period compared to an only 9 % drop for the 2020 pre-lockdown period, both compared to 2019. Meteorology accounted for SO₂ and CO concentrations show no significant changes during the 2020 lockdown period compared to 2019. GEOS-Chem (GC) simulation with a COVID-19 emission reduction scenario based on the observations (23 % reduction in anthropogenic NO_X emission with unchanged anthropogenic VOC and SO_2) are consistent with the small reductions of $PM_{2.5}$ during the lockdown and are used to identify the underlying drivers for this. Due to being in a NO_X saturated ozone production regime, GC OH radical and O₃ concentrations increased (15 and 9 %, respectively) during the lockdown compared to a Business As Usual (no lockdown) scenario. O_X (=NO₂+O₃) analysis implies that the increase in ozone at night-time is solely due to reduced NO titration. The increased O₃ results in increased NO₃ radical concentrations, primarily during the night, despite the large reductions in NO₂. Thus, the oxidative capacity of the atmosphere is increased in all three important oxidants, OH, O₃, and NO₃. PM nitrate formation from gas-phase nitric acid (HNO₃) is decreased during the lockdown as the increased OH concentration cannot compensate for the strong reductions in NO₂ resulting in decreased day-time HNO₃ formation from the OH + NO₂ reaction. However, night-time formation of PM nitrate from N₂O₅ hydrolysis is relatively unchanged. This results from the fact that increased night-time O₃ results in significantly increased NO₃ which roughly balances the effect of the strong NO₂ reductions on N₂O₅ formation. Ultimately, the only small observed decrease in lockdown PM_{2.5} concentrations can be explained by the large contribution of night-time PM nitrate formation, generally enhanced sulfate formation and slightly decreased ammonium. This study also suggests that high PM_{2.5} episodes in early spring are linked to high atmospheric ammonia concentrations combined with favorable meteorological conditions of low temperature and low boundary layer height. North-West Germany is a hot-spot of NH₃ emissions, primarily emitted from livestock farming and intensive agricultural activities (fertilizer application), with high NH₃ concentrations in the early spring and summer months. Based on our findings, we suggest that appropriate NO_X and VOC emission controls are required to limit ozone, and that should also

¹Environmental Sensing and Modeling, Technical University of Munich (TUM), Munich, Germany

²School of Engineering and Applied Science, Harvard University, Cambridge, MA, USA

³Department of Chemistry and Chemical Biology, Harvard University, Cambridge, MA, USA

help reduce $PM_{2.5}$. Regulation of NH_3 emissions, primarily from agricultural sectors, could result in significant reductions in $PM_{2.5}$ pollution.

1 Introduction

To halt the spread of the COVID-19 virus, various strict measures such as social isolation, curfews, and travel restrictions were implemented around the world in early 2020 (Steinmetz et al., 2020). As a result of these restrictions, anthropogenic emissions decreased significantly (Schumann et al., 2021; Le Quéré et al., 2020; Turner et al., 2020). Reduced primary emission activities from road transportation and industrial activities were expected to improve air quality. Numerous studies using satellite and insitu measurements have reported significant reductions in primary air pollutant concentrations during the COVID-19 lockdown period compared to pre-lockdown period in various parts of the world (Bauwens et al., 2020; Biswal et al., 2020; Collivignarelli et al., 2020; Dietrich et al., 2021; Field et al., 2020; He et al., 2021; Pathakoti et al., 2020; Mendez-Espinosa et al., 2020), but also emphasize the importance of accounting for the effects of different meteorological conditions between the study period and the reference period (Barré et al., 2020; Grange et al., 2020; Kroll et al., 2020; Koukouli et al., 2021; Ordóñez et al., 2020; Solberg et al., 2021). Anomalies in air pollutant concentrations caused by changes in meteorological conditions were also separated from observed changes using modeling work to estimate the actual influence of COVID-19 lockdown restrictions on air pollutant concentration changes (Balamurugan et al., 2021; Goldberg et al., 2020; Kang et al., 2020; Petetin et al., 2020; Qu et al., 2021; Yin et al., 2021). Secondary pollutant concentrations (O₃ and PM_{2.5}), which are primarily produced by precursor gases through complex atmospheric chemical reactions, remarkably increased or did not reduce commensurate to precursor emission reductions seen in some parts of the world during the COVID-19 lockdown period (Campbell et al., 2021; Deroubaix et al., 2021; He et al., 2021; Huang et al., 2021; Keller et al., 2021; Lee et al., 2020; Putaud et al., 2021; Souri et al., 2021; Wang et al., 2020, 2021).

Particulate Matter (PM) is the sum of all particles (solid and liquid) suspended in air, and can be classified based on aero-dynamic behavior, i.e., aerodynamic diameter (AD). Particles with an AD smaller than $10 \mu m$ are referred to as PM_{10} , while particles smaller than $2.5 \mu m$ AD are referred to as $PM_{2.5}$. Understanding of seasonal and inter-annual variability of PM, particularly over urban areas, remains a challenge (Fuzzi et al., 2015). This is mainly due to a lack of understanding in the attribution of PM sources. PM sources include both direct/primary sources (vehicle and industrial emissions, wind-blown dust, pollen, wildfires, etc.) as well as secondary formation (gas-to-particle conversion process) via atmospheric chemical reaction of precursor compounds such as NO_X (nitrogen oxides), SO_2 (sulfur dioxide), NH_3 (ammonia), VOCs (Volatile Organic Compounds) and other organic compounds, including compounds that have partitioned from primary aerosol back to the gas-phase, followed by partitioning to the condensed phase (Allen et al., 2015; Ayres et al., 2015; Fisher et al., 2016; Hallquist et al., 2009; Jacob, 1999; Jacobson, 1999; Marais et al., 2016; Seinfeld and Pankow, 2003; Steinfeld, 1998; Zhang et al., 2015). The composition of PM thus varies greatly depending on time and location; for example, in urban areas nitrate and organic aerosol often dominate in winter time Cesari et al., 2018; Zhai et al., 2021 (Cesari et al., 2018; Juda-Rezler et al., 2020; Samek et al., 2020; Salameh et al., 2015; Womack et al., 2019; Zhai et al., 2021).

In this study, we mainly focus on the response of urban surface PM_{2.5} to COVID-19 lockdown restrictions in Germany. Because major anthropogenic emissions are reduced, this unplanned intervention can test the understanding of the contribution of secondary PM_{2.5} sources, as well as the processes important in secondary PM_{2.5} formation. Despite of significant reductions in some anthropogenic activities, natural and agricultural air pollutant sources were not affected by the COVID-19 lockdown measures. Ammonia (NH₃) emissions (agricultural sources) are a significant source of PM_{2.5} in Germany in the spring (Fortems-Cheiney et al., 2016), when lockdown restrictions are implemented. Secondary inorganic aerosols such as ammonium sulfate and ammonium nitrate are the largest contributors to PM_{2.5} in Europe (Pay et al., 2012; Petetin et al., 2016). In comparison to sulfate formation, nitrate formation is more dependent on NH₃ concentration (Erisman and Schaap, 2004; Sharma et al., 2007; Wu et al., 2008). In the winter and spring (low temperature and high relative humidity), the role of NH₃ in PM_{2.5} formation is greater than in the summer (high temperature and low relative humidity) (Schiferl et al., 2016; Squizzato et al., 2013; Viatte et al., 2020). Primary components of PM_{2.5} are directly proportional to primary emission but secondary components of PM_{2.5} are not directly proportional to secondary precursor emissions or concentrations as they are produced by non-linear complex atmospheric chemical reactions (Shah et al., 2018). Observational and modeling evidence is required to estimate the influence of change in precursor emissions on $PM_{2.5}$ concentrations. To this end, we used ground and space-based measurements of PM_{2.5}, NO₂, O₃, SO₂, CO and NH₃ in conjunction with GEOS-Chem simulations to investigate the influence of lockdown restrictions on PM_{2.5} concentrations.

Modelling studies such as Gaubert et al. (2021); Hammer et al. (2021); Matthias et al. (2021); Menut et al. (2020) have already reported the $PM_{2.5}$ changes across Europe including Germany, during the COVID-19 lockdown period. The activity data (e.g., transportation, industrial activities and energy production) were used in the above mentioned studies to create a COVID-19 emission reduction scenario (Doumbia et al., 2021; Guevara et al., 2021). However, there are large discrepancies between various activity data sets (Gensheimer et al., 2021), necessitating different approaches to estimating the actual emission reduction caused by the COVID-19 lockdown restrictions. In this study, GEOS-Chem simulations (using identical anthropogenic emission for 2020 and 2019) were used to estimate the meteorology accounted for observed pollutant concentrations changes between 2020 and 2019, which were then used as a proxy for emissions reductions caused by COVID-19 lockdown measures to create a COVID-19 emission scenario in GEOS-Chem model for simulating the lockdown pollutant concentrations (Fig. 1). In addition to looking at the impact of lockdown restrictions on air pollutant concentrations (Sect. 4.1), we focus on process level analysis of the impact of changes in precursor emissions (NO $_X$) on PM $_{2.5}$ formation (Sect. 4.2), as well as the role of ammonia (NH $_3$) emissions in PM $_{2.5}$ formation (Sect. 4.3).

2 Data and Model

75

85

Data sets used in this study are summarized in Table 1. We focused on ten metropolitan areas in Germany (Bremen, Cologne, Dresden, Dusseldorf, Frankfurt, Hamburg, Hanover, Leipzig, Munich and Stuttgart) and used surface air pollutant concentration data (PM_{2.5}, NO₂, O₃) for all of these while SO₂ data was only available for five of these areas (Bremen, Dresden, Frankfurt, Hamburg and Leipzig) and CO data was limited to six metropolitan areas (Bremen, Frankfurt, Hamburg, Hanover, Munich

Table 1. Data sets used in this study.

| Data source | Data | Temporal resolution | Spatial resolution | Data availability |
|--|---|---------------------|---|---|
| Governmental | NO ₂ , O ₃ , PM _{2.5} | 1 h | - | Bremen, Cologne, Dresden, Dusseldorf, Frankfurt, Hamburg, Hanover, Leipzig, Munich and Stuttgart metropolitan areas |
| in-situ measurements | SO_2 | 1 h | - | Bremen, Dresden, Frankfurt, Hamburg and Leipzig metropolitan areas |
| | СО | 1 h | - | Bremen, Frankfurt, Hamburg, Hanover, Munich and Stuttgart metropolitan areas |
| TROPOMI satellite measurements | SO_2 | daily | 7*3.5 km (5.5*3.5 km, after August 6, 2019) | All of Germany |
| IASI satellite | NH ₃ | twice a day | 12 km diameter | All of Germany |
| measurements | | monthly | 1 degree | All of Germany |
| ERA 5 (ECMWF reanalysis) | Temperature, relative humidity, boundary layer height and wind speed | 1 h | 0.25 degree | All of Germany |
| | Precipitation | daily | 1 degree | All of Germany |
| GEOS-Chem (GC) chemical transport model | All species | 1 h | 0.5 * 0.625 degree | All of Germany |

and Stuttgart). We use data for 2019 and 2020 in this work (data-obtained from https://discomap.eea.europa.eu/map/fme/AirQualityExport.htm).

TROPOMI tropospheric SO_2 (Theys et al., 2017) column products are also used (offline products-obtained from https: //s5phub.copernicus.eu). The TROPOMI SO_2 product provides the total SO_2 column between the surface and the top of troposphere. The TROPOMI overpass occurs around 13.30 local time. At the start of the mission, the TROPOMI product provided data at a resolution of 7*3.5 km, while after August 6, 2019 the resolution improved to 5.5*3.5 km. Stricter quality filtering criteria (quality assurance value (qa) >= 0.5) was applied to the dataset. A daily mean of SO_2 is calculated by averaging these values within 0.5-degree radius of the urban center.

The daily atmospheric NH₃ variability in Germany was studied using the "near-real time daily IASI/Metop-B ammonia (NH3) total column (ANNI-NH3-v3)" dataset (products-obtained from https://iasi.aeris-data.fr/catalog/). The data used are from the IASI instrument aboard the Metop-B satellite, which has a local solar overpass time of 9:30 a.m and 9:30 p.m (Clerbaux et al., 2009). We only used day-time (9.30 am) measurements in this study. Night-time measurements (9.30 pm) were excluded due to their large relative errors. A daily mean is calculated by averaging the values within 0.5-degree radius of the urban center. The monthly atmospheric NH₃ variability in Germany was studied using the "standard monthly IASI/Metop-B ULB-LATMOS ammonia (NH3) L3 product (total column)" dataset. This product contains a monthly averaged NH₃ total column with a spatial resolution of 1*1 degree (products-obtained from https://iasi.aeris-data.fr/catalog/).

Temperature, relative humidity, boundary layer height and wind information are obtained from the ERA 5 product (Hersbach et al., 2020). This product's native spatial and temporal resolutions are 0.25 degree and 1 hour, respectively. For precipitation information, the GPCP daily gridded product from ERA 5 is used, which provides global gridded data at 1-degree resolution (products-obtained from https://cds.climate.copernicus.eu/).

We used the GEOS-Chem (GC) chemical transport model (http://doi.org/10.5281/zenodo.3959279) to simulate the pollutant concentration for 2020 and 2019. The GC simulation conducted over Germany (4-17°E, 45-57°N) had a horizontal resolution

of 0.5°*0.625° with dynamic boundary conditions generated from a global simulation with 4°*5° resolution. We ran the GC simulation for two cases. In the first case, anthropogenic emissions from the 2014 CEDS inventory (Hoesly et al., 2018), the most recent version of which is 2014, are used in the GC simulations for both 2019 and 2020, but with the corresponding meteorology from MERRA-2 global reanalysis product for 2019 and 2020. Natural emissions from soil and lightning are calculated for the corresponding year using mechanisms described in Hudman et al. (2012) and Murray (2016). The corresponding year's open fire emissions from GFED4 (Werf et al., 2017) are used for 2019 and 2020. In the second case, the 2014 CEDS anthropogenic emission inventory were scaled down by the estimated emissions reduction caused by the lockdown restrictions for the 2020 lockdown period. The remaining (natural and fire) emissions are calculated in the same way as in the first case. Even though the 2014 CEDS anthropogenic emission inventory is used in GC simulations, the effects of anthropogenic emission changes between 2014 and 2019 or 2020 will be canceled out because we use the difference between two years (e.g., 2020-2019) or two cases (e.g., 2020_{lockdown} - 2020_{no lockdown}) in our study.

125 3 Method

130

115

The following is our methodology for estimating meteorology accounted for observed pollutant concentration changes between 2020 and 2019, similar to Balamurugan et al. (2021); Qu et al. (2021). We estimate the difference in pollutant concentrations between 2020 and 2019 caused by changes in meteorology using GC simulated concentrations (first case). Since GC uses identical anthropogenic emission for 2020 and 2019, with the corresponding year meteorology, the difference between 2020 and 2019 GC pollutant (e.g., $PM_{2.5}$) concentrations only results from meteorology changes between 2020 and 2019. We use Δ to signify absolute concentration change, and f to signify fractional (percentage) change.

$$\Delta P M_{2.5(GC)} = P M_{2.5(GC,2020)} - P M_{2.5(GC,2019)} \tag{1}$$

The observed (ground-truth measurements) pollutant concentration changes between 2020 and 2019, which includes the effects of lockdown restrictions and meteorology, is:

135
$$\Delta PM_{2.5(obs)} = PM_{2.5(obs,2020)} - PM_{2.5(obs,2019)}$$
 (2)

To disentangle the meteorology contribution from the observed pollutant concentration changes, we subtract the GC pollutant concentration changes caused by meteorology from observed pollutant concentration changes between 2020 and 2019.

$$\Delta P M_{2.5(obs,emi)} = \Delta P M_{2.5(obs)} - \Delta P M_{2.5(GC)} \tag{3}$$

The fractional change in meteorology accounted for pollutant concentration between 2020 and 2019, i.e. pollutant concentration

thanges between 2020 and 2019 due to emission changes only fractional change (%) in pollutant concentration between 2020 and 2019 due to emission changes only, is calculated as,

$$fPM_{2.5(obs,emi)} = \frac{\Delta PM_{2.5(obs,emi)}}{PM_{2.5(obs,2019)}} *100 \tag{4}$$

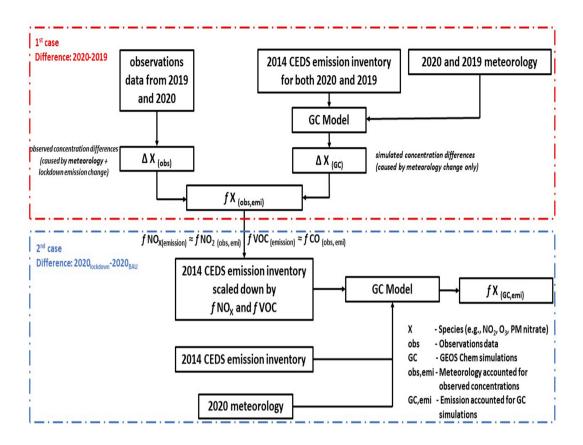


Figure 1. Schematic diagram of our methodology for calculating the meteorology accounted for observed pollutant concentrations changes between 2020 and 2019, and emission accounted for GC pollutant concentrations changes between 2020 lockdown and 2020 BAU scenario.

where, "obs", "GC" and "obs,emi" refer to ground-truth measurements (observations data), GEOS-Chem simulations and meteorology accounted for ground-truth measurements, respectively.

145

150

We estimate the meteorology accounted for fractional change in other pollutant concentrations analogously. Our previous study (Balamurugan et al., 2021), using the same methodology, reported the meteorology accounted for NO_2 and O_3 concentration changes for eight German metropolitan areas. Here, we reproduce the results for NO_2 and O_3 concentrations, but for ten metropolitan areas. We use $fNO_{2(obs,emi)}$ and $fCO_{(obs,emi)}$ to capture fractional changes in anthropogenic NO_X and VOC emission $(fNO_{X(emission)})$ and $fVOC_{(emission)})$ due to lock down restrictions, respectively. Because of the scarcity of VOC measurements, CO data was used as a proxy for anthropogenic VOC (Fujita et al., 2003; Jiménez et al., 2005; Stephens et al., 2008; Yarwood et al., 2003) and NO_2 was used as proxy for NO_X . This assumption is supported by studies such as Baker et al. (2008); Von Schneidemesser et al. (2010), which show anthropogenic VOC is well correlated with CO, and Blanchard and Tanenbaum (2003), which shows comparable changes in VOC and CO between weekday and weekend. Changes in biogenic

VOCs are not directly affected by lockdown measures.

$$155 \quad fNO_{X(emission)} \approx fNO_{2(obs,emi)} \tag{5}$$

$$fVOC_{(emission)} \approx fCO_{(obs.emi)}$$
 (6)

The base anthropogenic emission inventory were then scaled down by $fNO_{X(emission)}$ and $fVOC_{(emission)}$ for NO_X and VOC emission, respectively, in the GC model for the 2020 lockdown period (second case), which simulates all pollutants concentrations for the lockdown emission scenario. The fractional change in emission accounted for, i.e. using scaled emission inventories, GC pollutants level during the 2020 lockdown period compared to 2020 Business As Usual (BAU), i.e., no lockdown, level is calculated as,

$$fPM_{2.5(GC,emi)} = \frac{PM_{2.5(GC,2020,lock)} - PM_{2.5(GC,2020)}}{PM_{2.5(GC,2020)}} *100$$
(7)

where, "GC,emi" refers to GC simulations accounting for scaled emission and PM_{2.5(GC,2020,lock)} are the PM_{2.5} concentrations during the lockdown period determined via the 2020 GC simulations with down-scaled emissions. We estimate the emission accounted for concentration changes of other pollutants in the same way. Figure 1 illustrates our methodology for calculating the meteorology accounted for observed pollutant concentrations changes between 2020 and 2019, as well as emission accounted for GC pollutant concentration changes between 2020 lockdown and 2020 BAU scenario.

4 Results and discussion

160

180

170 4.1 Influence of lockdown restrictions on the concentrations of air pollutants

To assess the impact of lockdown restrictions on the concentration of air pollutants, we compared the 2020 lockdown period pollutant concentrations to the same period in 2019. These comparison results, however, need to take the effects of both meteorological and lockdown restrictions into account. As mentioned in Sect. 3, we used GEOS-Chem simulations to disentangle the effects of meteorology on observed pollutant concentration changes between 2020 and 2019. Studies such as Balamurugan et al. (2021) and Tai et al. (2012) have shown that GEOS-Chem can reproduce the temporal variability of observed pollutant concentrations including PM_{2.5}, emphasizing that GC can be used for process level analysis of PM_{2.5} variability. We also compared the 2019 GC and 2019 observed in-situ PM_{2.5} concentrations and found that the GC and observed in-situ PM_{2.5} concentrations were in good agreement (R > 0.5 for all metropolitan areas, except Leipzig which has a R value of 0.39) (e.g., Fig. 6 (e), for Cologne metropolitan area). Table A1 shows the statistical evaluation (R, RMSE and mean bias) of the GC model performance for each metropolitan area. The GC simulations underestimate the PM_{2.5} when compared to observed in-situ PM_{2.5} concentrations (mean bias) ((GC - in-situ)/in-situ) ranges from -12.7 % to -37.4 %), except for the Cologne metropolitan area (+ 11.7 %). However, since we use the GC's relative difference between 2020 and 2019, this bias should cancel out. We also compared the 2019 GC simulated nitrate and ammonium concentration for the urban measurement station in Germany (14.33°E,

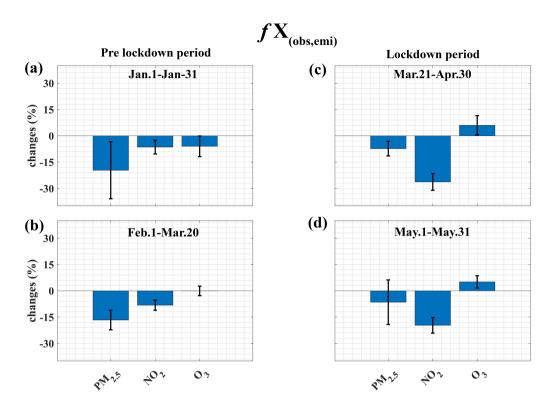


Figure 2. Meteorology accounted for mean in-situ PM_{2.5}, NO₂, and O₃ concentration changes between 2020 and 2019. Results of computations according to our first case $(fX_{(obs,emi)})$ in the Sect. 3. Error bars represent the 1 σ of mean of ten metropolitan areas.

51.75°N). The statistical evaluation (R, RMSE and mean bias) of the GC model performance for these species is given in Table

B1.

Figure 2 shows meteorology accounted for mean PM_{2.5}, NO₂ and O₃ concentration changes between 2020 and 2019 for ten German metropolitan areas from January 1 through May 31. Both meteorology accounted and unaccounted for mean PM_{2.5}, NO₂ and O₃ concentration changes between 2020 and 2019 for ten German metropolitan areas are shown in Appendix Fig. A1. The German government imposed COVID-19 lockdown restrictions on March 21, 2020 in Germany. In figures and for specific cases, the pre-lockdown period (January 1 to March 20) is divided into two sections, and the lockdown period (March 21 to May 31) is also divided into two sections (unless otherwise specified): (a) January 1 to January 31, 2020 - No lockdown restrictions, (b) February 1 to March 20, 2020 - No lockdown restrictions in the event of unusual weather conditions (occurrence of storms), (c) March 21 to April 30, 2020 (spring) - Strict lockdown measures, and (d) May 1 to May 31, 2020 (late spring) - Loose lockdown measures. Germany experienced high wind conditions due to storms in February 2020 (Matthias et al., 2021), which was used to determine the extent of meteorology's role in pollutant concentration changes. Meteorology unaccounted for mean NO₂ and PM_{2.5} concentrations for February 1 to March 20, 2020 period (before the implementation of lockdown) are lower than the corresponding ones in 2019 by 30 % and 42 % (fNO_{2(obs)}) and fPM_{2.5(obs)}), respectively, due to

the dilution/dispersion from the high wind conditions. However, after accounting for meteorology, the difference in mean NO_2 and $PM_{2.5}$ concentrations between 2020 and 2019 for the period February 1 to March 20 ($fNO_{2(obs,emi)}$) and $fPM_{2.5(obs,emi)}$) are 8 % and 18 %, respectively. This finding is consistent with meteorology accounted for mean NO_2 and $PM_{2.5}$ changes between 2020 and 2019 for the period January 1 to January 31 (Fig. 2 (a,b)). This highlights the importance of accounting for meteorological impacts.

200

205

210

220

225

230

In the 2020 pre-lockdown period (January 1 to March 20), both meteorology accounted for mean NO₂ and PM_{2.5} levels are lower by 9 % and 19 %, respectively, compared to the same period in 2019. During the 2020 lockdown period (March 21 to May 31), mean meteorology accounted for NO₂ concentrations dropped significantly (23 %) compared to the same period in 2019, which is greater than the drop in the 2020 pre-lockdown period compared to 2019 (9 %). Comparatively, mean meteorology accounted for 2020 lockdown PM_{2.5} concentrations show a smaller reduction (5 %) compared to the same period in 2019, while an important precursor, NO₂, decreased by 23 % during the same period. Furthermore, the meteorology accounted for PM_{2.5} reduction during the 2020 lockdown period (5 %) is less than the meteorology accounted for PM_{2.5} reduction observed during the 2020 pre-lockdown period (19 %) compared to the corresponding 2019 periods (Fig. 2). Especially in Munich and Stuttgart, meteorology accounted for PM_{2.5} concentrations during the 2020 lockdown period are higher than in 2019. The meteorology accounted for mean O3 concentrations in the 2020 lockdown period are increased by 6 % compared to the same period in 2019. The increase in O_3 concentration during the 2020 lockdown period is mainly due to being in a NO_X saturated regime (Gaubert et al., 2021), in which reducing NO_X emission results in an increase in O_3 concentrations (Sillman, 1999; Sillman et al., 1990). It is also possible that the increase in ozone is due to less ozone destruction via lower NO titration, in addition to an increase in ozone formation efficiency through NO_X saturated regime chemistry. The meteorology accounted for mean O_X (= $NO_2 + O_3$) concentrations in the 2020 lockdown period is 2 % higher than in 2019 (Fig. C1(a)), implies that the reduced NO titration effect partly contributed to the increased ozone. O_X analysis also implies that the decrease in NO_2 was offset by an increase in O_3 , and ozone production is overwhelmingly NO_X saturated in Germany.

The effects of lockdown restrictions on SO₂ concentrations are insignificant. In comparison to 2019, TROPOMI meteorology accounted for SO₂ levels are decreased by 1 % during the 2020 lockdown period compared to 2019 (Fig. A1). For accounting meteorological impacts on TROPOMI satellite column concentrations, GEOS-Chem diagnostics (47 vertical layers) were converted to a column, applying TROPOMI's averaging kernel. Because of the large influence of background concentration on satellite column measurements, we also investigated in-situ SO₂ concentrations, but only for five metropolitan areas. Similarly, we found that the impact of lockdown restrictions on in-situ SO₂ concentrations is marginal (Fig. B1). The road transportation sector contributes less than 1 % of total sulfur dioxide emissions, while coal-related fuel burning (industrial and energy production) accounts for nearly 80 % of total sulfur dioxide emissions (SO2, 2021). Because the lockdown restrictions primarily reduced traffic-related emissions, we see far less effects of the lockdown on SO₂ concentration (slight increase or no significant decrease in other European metropolitan areas (Collivignarelli et al., 2020; Filonchyk et al., 2021; Higham et al., 2021)). We found similar effects on in-situ CO concentration changes in six metropolitan areas. The meteorology accounted for mean CO concentrations are lower by 3 % during the 2020 lockdown period compared to 2019 (Fig. B1). Stuttgart meteorology

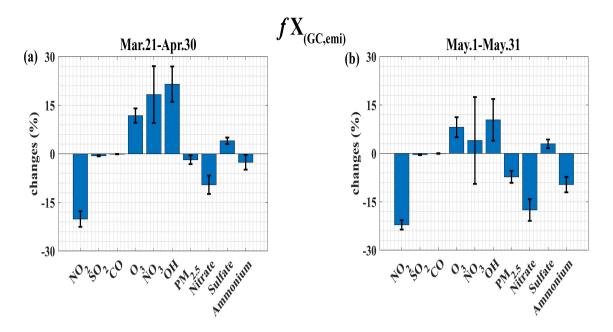


Figure 3. The emission accounted for GC NO₂, SO₂, CO, O₃, NO₃ radical, OH radical, PM_{2.5}, inorganic nitrate, sulfate, ammonium concentration changes between 2020 lockdown and 2020 BAU (no lockdown) scenario ($fX_{(GC,emi)}$). Error bars represent the 1 σ of mean of ten metropolitan areas.

accounted for CO concentrations in 2020 were higher than 2019 at all times. Other metropolitan areas experienced minor reductions (Clark et al., 2021; Hörmann et al., 2021).

4.2 Model evidence of changes in air pollutants concentration resulting from lockdown restrictions

235

240

245

As mentioned in Sect. 3, we use the meteorology accounted for NO_2 and CO changes to adjust the anthropogenic NO_X and VOC emissions in inventories due to lockdown restriction impacts. GC model simulations are then obtained with this scaled anthropogenic emission scenario (23 % reduction in anthropogenic NO_X emission and unchanged anthropogenic VOC emissions) for the 2020 lockdown period. The NO_X emission reduction is within the range of estimated anthropogenic NO_X emission reductions using activity data for Europe by previous authors (Doumbia et al., 2021; Guevara et al., 2021) (25 % and 33 %, respectively). For those studies there are large differences in estimated anthropogenic VOC emission changes for Europe; Doumbia et al. (2021) estimated 34 % while Guevara et al. (2021) estimated 8 % reduction in anthropogenic VOC emissions, compared to the BAU scenario. However, the real-time measurements at a United Kingdom station show no significant changes in many VOC concentrations during the lockdown period (Grange et al., 2020). For the NO_X saturated ozone production regime regime, VOC emission reductions can decrease ozone levels, while NO_X emission reductions increase them. Gaubert et al. (2021) conducted a sensitivity study of modelling work on ozone levels in response to the NO_X or VOC or both emission reductions for the 2020 lockdown period. The reduction in both emissions (NO_X and VOC), suggested by Doumbia et al. (2021),

results in slight increase in lockdown ozone levels (< 2.5 %) over only north-western Germany and slight decrease in lockdown ozone levels over other regions of Germany, compared to BAU levels. But, only reduction in NO_X emission results in increased lockdown ozone levels (0-10 %) over all of Germany compared to BAU levels, which is also consistent with our results of increase in meteorology accounted for ozone levels over different metropolitan areas across Germany during 2020 lockdown period compared to 2019 levels. This implies that anthropogenic VOC emissions were either not reduced at all or by a much smaller percentage than anthropogenic NO_X emissions, compared to the BAU scenario. According to the European Environment Agency (EEA) (https://www.eea.europa.eu/data-and-maps/indicators/eea-32-non-methane-volatile-1/assessment-4), the road transport sector accounts for 14.6 % of total NMVOC emissions, while the road transport sector accounts for 40.5 % of total NO_X emissions (https://www.eea.europa.eu/data-and-maps/indicators/eea-32-nitrogen-oxides-nox-emissions-1/assessment.2010-08-19.0140149032-3: :text=EEA%2D33%20emissions%20of%20nitrogen,households'%20(13%25)%20sectors). According to Guevara et al. (2021), the transportation sector accounts for nearly 90 % of the reduction in total anthropogenic NO_X and VOC emissions during lockdown. As we noted that NO_X emission decreased by 23 %, and the lockdown restrictions primarily reduced traffic-related emissions, we can directly extrapolate this to a reduction in road transportation-related emissions; approximately 43 % (23-40.50 / 40.50). This finding also corresponds to a 40 % decrease in traffic vehicle count (Gensheimer et al., 2021). Therefore, the decrease in VOC emission from transport sector should be 6 % (14.6 * 0.43). However, due to a significant decline in the transport sector's VOC emission in recent years, this reduction in VOC emission from the transport sector, calculated based on the EEA's 2015 data, should be even less than 6 %. There is also no evidence that lockdown measures affect the major source of VOC emissions, which are use of volatile chemical products such as cleaning agents and personal care products, as well as biogenic emissions.

250

255

260

265

275

280

The emission accounted for GC lockdown NO_2 concentrations decreased by 21 % ($fNO_{2(GC,emi)}$) while emission accounted for GC lockdown O_3 concentrations increased by 9 % compared to 2020 BAU (Fig. 3). This is consistent with previous studies (such as Balamurugan et al. (2021); Gaubert et al. (2021)) which show that German metropolitan areas are in a NO_X saturated ozone production regime in spring. However, the diurnal cycle of GC O_X changes between 2020 lockdown and BAU suggests that night-time ozone increases are solely due to a decrease in NO titration effects (Fig. C1(b)). The emission accounted for GC lockdown PM concentrations show small decreases compared to 2020 BAU (Fig. 3). These results are consistent with previous studies (Gaubert et al., 2021; Hammer et al., 2021; Matthias et al., 2021; Menut et al., 2020), which used activity data to develop an emission reduction scenario and estimated small to no reduction in $PM_{2.5}$, a significant drop in NO_2 and marginal increase in O_3 levels during 2020 lockdown period, compared to BAU levels, over Northern-Europe including Germany.

We investigated the GC $PM_{2.5}$ composition for the studied period to determine the role of reduced NO_X emission on total $PM_{2.5}$. Major secondary $PM_{2.5}$ components are nitrate, sulfate, ammonium and organic aerosol, which, on average, correspond to 24 %, 23 %, 15 % and 30 % of total $PM_{2.5}$, respectively, during March 21 to May 31, 2019 (Fig. D1). Mean relative contribution of $PM_{2.5}$ species for 2020 (BAU) and 2020 (lockdown) are shown in Fig. E1 and F1, respectively. The emission accounted for GC PM nitrate levels during the 2020 initial lockdown period (March 21 to April 30) are 9.5 % lower than the 2020 BAU levels ($fNIT_{(GC,emi)}$) (Fig. 3 (a)), however, we see NO_2 decreased by 21 % during the same period.

The decrease in emission accounted for GC PM nitrate is also less than the decrease in NO₂ during the second half of the lockdown (May 1 to May 31). The emission accounted for GC lockdown PM sulfate level show marginal increase (3.5 %), while emission accounted for GC lockdown PM ammonium shows marginal decrease (5.8 %), compared to 2020 BAU level. The slight increase (& decrease) in sulfate (& ammonium) was also found in the Hammer et al. (2021); Matthias et al. (2021) studies, which used activity data to adjust the COVID-19 emission scenario.

It is notable that the reduction in NO_X , a precursor to PM nitrate, does not directly translated into a decrease in PM nitrate formation. There are several pathways for the formation of nitric acid (HNO₃), which partition to PM nitrate (Allen et al., 2015; Bauer et al., 2007). The reaction of OH and NO_2 (homogeneous pathway) and the hydrolysis of N_2O_5 on aerosol particles (heterogeneous pathway) are the two major pathways (Chang et al., 2011, 2016; Mollner et al., 2010).

The reaction for HNO₃ formation via gas-phase oxidation of NO₂ by OH is:

$$NO_2 + OH \xrightarrow{M} HNO_3$$
 (R1)

The reactions resulting in HNO₃ formation via hydrolysis of N₂O₅ on aerosol surfaces are:

$$NO_2 + O_3 \longrightarrow NO_3 + O_2$$
 (R2)

295

285

290

$$NO_3 + NO_2 \stackrel{M}{\longleftrightarrow} N_2O_5$$
 (R3)

$$N_2O_5 + H_2O(1) \longrightarrow 2 HNO_3$$
 (R4)

The formation of HNO₃ from the reaction of OH and NO₂ dominates during the day, while hydrolysis of N₂O₅ on aerosol particles dominates at night as OH night-time concentrations are low and N₂O₅ photolyzes easily (Russell et al., 1986). At night, NO₃ radical can be an important precursor for PM nitrate via reactions (Eq. R3, R4) (Kang et al., 2021; Shah et al., 2020; Wang et al., 2013). The emission accounted for concentrations of OH and NO₃, which drive day and night-time formation of PM nitrate, increased substantially (15 % and 12 %, respectively) during the lockdown period compared to BAU (Fig. 3). The increase in OH radicals results from German metropolitan areas being in a NO_X saturated regime (Shah et al., 2020). The increase in GC lockdown NO₃ levels is predominantly at night due to a significant increase in night-time O₃ (Fig. 4 (b,e)); the reaction of NO₂ with O₃ is the most important source of NO₃ radical (Eq. R2) (Geyer et al., 2001).

Liu et al. (2020) have demonstrated that analyzing the diurnal cycle of total inorganic nitrate helps to identify the dominant pathway for the particulate nitrate production. The emission accounted for GC lockdown PM nitrate levels decreased significantly during the day, while night-time lockdown PM nitrate levels decreased slightly compared to BAU levels (Fig. 4 (h)). Even though GC lockdown OH levels increased, HNO₃ production from the OH+NO₂ reaction during the lockdown period is reduced due to significantly lower day-time NO₂ levels compared to BAU (Fig. 4 (d)); as a result, GC day-time lockdown PM nitrate levels are significantly lower compared to BAU levels. However, higher night-time NO₃ levels result in higher night-time

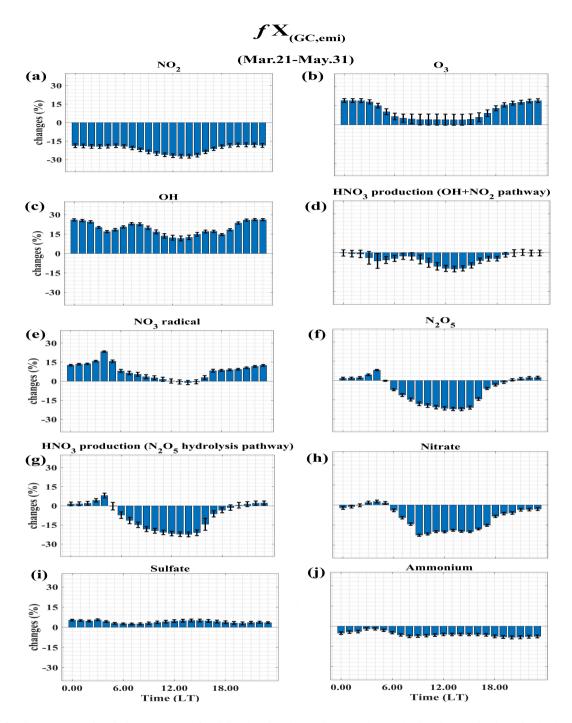


Figure 4. Diurnal cycle of emission accounted for GC NO₂, O₃, OH radical, HNO₃ production from oxidation of NO₂ by OH pathway, NO₃ radical, N₂O₅, HNO₃ production from N₂O₅ hydrolysis pathway, PM nitrate, sulfate, ammonium concentration changes between 2020 lockdown and 2020 BAU (no lockdown) scenario ($fX_{(GC,emi)}$). Error bars represent the standard error of respective hour in ten metropolitan areas.

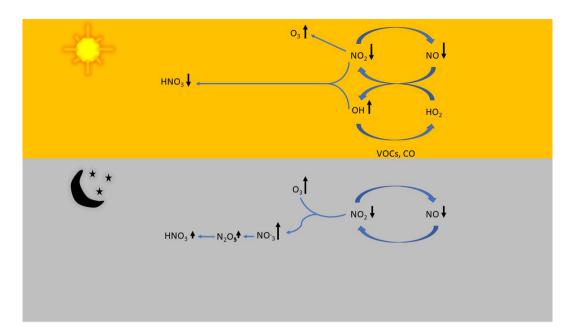


Figure 5. Generalized schematic diagram of day and night-time lockdown NO_X chemistry compared to BAU scenario.

HNO₃ production from N₂O₅ hydrolysis, resulting in slightly lower night-time lockdown PM nitrate compared to BAU (Fig. 4 (b,e,f,g)) However, higher night-time NO₃ levels result in relatively unchanged night-time HNO₃ production from N₂O₅ hydrolysis, resulting in slightly lower night-time lockdown PM nitrate compared to BAU (Fig. 4 (b,e,f,g)). This implies that the increase in NO₃ radical due to increased ozone partially offset the effect of reduced NO_X on nitrate formation. Previous studies have also shown that N₂O₅ hydrolysis plays important role in nitrate formation than the gas-phase day-time pathway (NO₂ + OH) (Allen et al., 2015; Chan et al., 2021; Kim et al., 2014; Liu et al., 2020; Yan et al., 2019). Figure 5 illustrates the conceptual model of generalized day and night-time lockdown NO_X chemistry compared to BAU scenario. The oxidation of SO₂ is a major source of sulfate, and the reaction with the OH radical dominates the gas-phase oxidation of SO₂ (Zhang et al., 2015). Therefore, the enhanced sulfate formation during the 2020 lockdown period could be due to the increased oxidizing capacity of atmosphere (OH) since we observe no significant change in emission accounted for GC SO₂ concentration, compared to BAU concentration (Fig. 3). Organic aerosol (OA) formation could be affected by the changes in oxidizing capacity of the atmosphere (Carlton et al., 2009), but no changes in emission accounted for GC lockdown OA were observed compared to 2020 BAU scenario. Therefore, the fact that no significant change in PM_{2.5} due to lockdown restrictions is observed can be explained by a significant offset of the decreased day-time PM nitrate formation by enhanced formation of PM sulfate, while PM ammonium shows a marginal decrease.

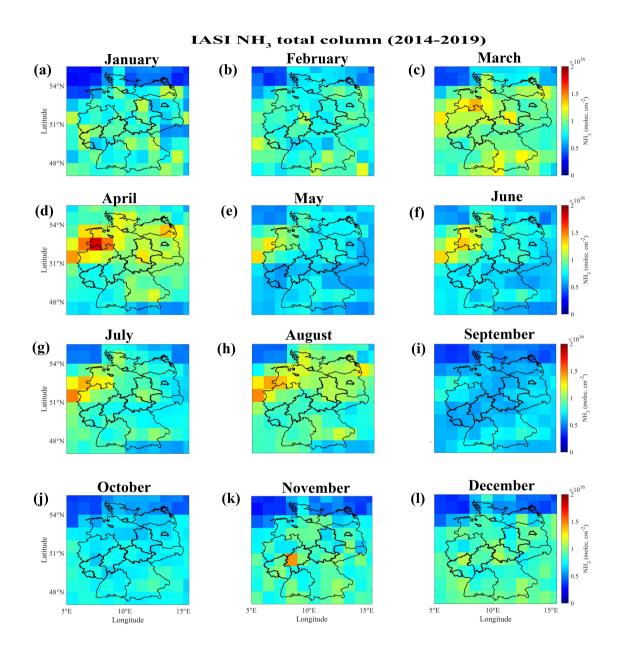


Figure 6. Monthly mean IASI NH₃ total column at 1*1 degree resolution.

4.3 Link between spring PM_{2.5} pollution episodes and high NH₃ concentrations

330

335

340

350

355

360

It is worth noting that a significant fraction of PM_{2.5} is PM nitrate. Ammonia (NH₃) is an important precursor for particulate nitrate formation (Ansari and Pandis, 1998; Banzhaf et al., 2013; Behera and Sharma, 2010; Wu et al., 2016). This explains the importance of monitoring and potentially regulating ammonia emissions. Therefore, the inter- and intra-annual changes in ammonia (NH₃) concentrations over Germany, as well as their relationship to PM_{2.5} variability, are reviewed and analyzed further below. In Germany, atmospheric NH₃ levels follow a monthly pattern, with NH₃ levels peaking in April (Fig. 6). NH₃ levels are also elevated during summer months. In Europe, major agricultural practices (fertilizer and manure applications) take place in the early spring (Petetin et al., 2016; Ramanantenasoa et al., 2018; Viatte et al., 2020). The higher atmospheric ammonia levels in April are attributable to agricultural practices such as fertilizer application. The high NH₃ values in summer are most likely due to warm climates (Kuttippurath et al., 2020). Monthly average NH₃ maps clearly show the high NH₃ values over North-West Germany from April to August, with particularly high values in April. It indicates that North-West Germany is a hotspot of ammonia emissions compared to the rest of the country. North-West Germany is known for its high livestock density (livestock farming (EUR, 2013; Scarlat et al., 2018)) and it is dominated by crop and grass land (ESA, 2017). Livestock farming and fertilizer application account for 75 % of NH₃ emissions in Europe (Webb et al., 2005). NH₃ concentrations in Germany vary greatly from year to year (inter-annual variabilities). We consider the period between March 21 and April 30 when a stricter lockdown was in place to illustrate the inter-annual variability of atmospheric NH₃ between 2018 and 2020 (Fig. 7). NH₃ levels are lower in 2019 than in 2018, which can be attributed to the lower temperature in 2019 compared to 2018. Meanwhile, even though strict lockdown was in place, NH₃ levels in 2020 are higher than in 2019 and 2018, possibly due to low precipitation. High temperatures promote NH₃ volatilization (increases the NH₃ level in the atmosphere) (Ernst and Massey, 1960), whereas high rainfall favors wet deposition (removal of atmospheric NH₃). Schiferl et al. (2016); Viatte et al. (2020) have also shown that meteorological parameters such as temperature and precipitation play a greater role in NH₃ inter-annual variability.

High PM pollution episodes are likely to occur frequently during the winter due to high residential heating demand and favorable meteorological conditions (e.g., low temperature and inversion condition). However, high concentrations of PM_{2.5} are apparent in German metropolitan areas in the early spring (from the second half of March to the end of April, e.g., Fig. 8 (a) for Cologne Munich metropolitan area). On March 21, 2020, the German government imposed COVID-19 lockdown restrictions. However, in-situ PM_{2.5} concentrations during the initial lockdown period are higher than during the pre-lockdown period in 2020. High PM_{2.5} levels from the second half of March to the end of April are also consistent with previous years without lockdown restrictions. It is notable that this high spring PM_{2.5} episodes are associated with high NH₃ concentrations (Fig. 6 (b)). The high PM_{2.5} events that occur in the spring have also been observed in other European cities, and they typically contain ammonium nitrate and ammonium sulfate (Fortems-Cheiney et al., 2016; Renner and Wolke, 2010; Schaap et al., 2004; Viatte et al., 2020, 2021). Above, we show the high NH₃ levels in early spring (April) and summer months. High PM_{2.5} concentrations are evident in spring, however, we did not observe high PM_{2.5} episodes in summer (Fig. 8 (a)). It is also worth noting that even in the spring and winter PM_{2.5} is not consistently high on days with high NH₃. This reflects the complexity of the process of gas to particle

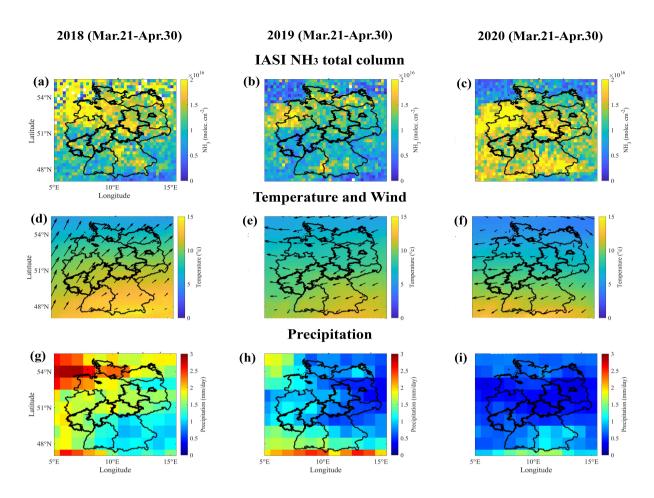


Figure 7. Mean IASI NH₃ total column (daily IASI NH₃ measurements gridded at 0.25 degree resolution) (top), mean temperature and wind (middle) and mean precipitation (bottom).

conversion. Despite high NH₃ concentrations, ammonia(NH₃)-to-ammonium(NH₄) conversion is mainly driven by various meteorological factors such as temperature (and relative humidity). Studies (Viatte et al., 2020; Wang et al., 2015; Watson et al., 1994) have shown that conditions such as temperature of less than 10 °C and a high relative humidity of more than 70 % are optimal for atmospheric gas-phase NH₃ to transform into ammonium salts, mainly due to reversible ammonium nitrate formation, which depends on temperature and relative humidity; warm and dry conditions partition ammonia back to the gas phase (Mozurkewich, 1993). In comparison to summer, the impact of NH₃ on PM_{2.5} formation is considerable for winter and spring over Europe (Viatte et al., 2020, 2021) and the US (Schiferl et al., 2016). Summer weather is typically warmer (and has lower relative humidity) than winter and spring, which could explain why high NH₃ concentrations are not associated with high PM_{2.5} in summer or late spring. Furthermore, it is important to note that PM_{2.5} anthropogenic precursor emissions (NO_X, SO₂, VOCs) have a seasonal cycle, with higher emissions in winter than summer; however, biogenic VOC emissions dominate in the summer. To further demonstrate this for German metropolitan areas To further demonstrate the relationship between PM_{2.5} and NH₃ for

365

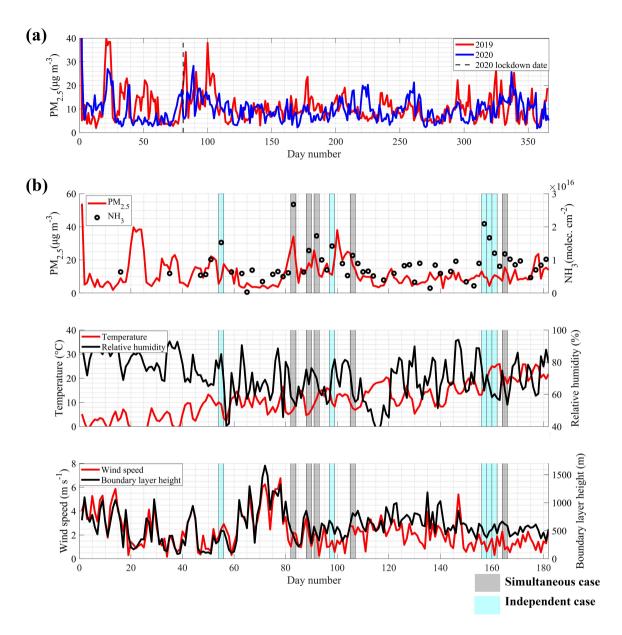


Figure 8. 2019 and 2020 annual daily mean in-situ $PM_{2.5}$ concentrations in Munich (a). In figure panel (a), the vertical dashed line denotes the start of 2020 lockdown. 2019 daily mean in-situ $PM_{2.5}$ and column NH_3 from IASI satellite (b, top). 2019 daily mean temperature and relative humidity (b, middle). 2019 daily mean wind speed and boundary layer height (b, bottom). The corresponding days for the cases "Simultaneous" are shaded with gray color and for the cases "Independent" are shaded with cyan color. "Simultaneous" - Simultaneous increase in NH_3 (IASI) and $PM_{2.5}$ (in-situ) concentrations on same day. "Independent" - Increase in NH_3 (IASI) concentration not corresponding to an increase in $PM_{2.5}$ (in-situ) concentration on same day.

German metropolitan areas, we consider two cases ("Simultaneous" and "Independent") for 2018 and 2019 (e.g., Fig. 8 (b) for Munich metropolitan area). "Simultaneous" - Simultaneous increase in NH₃ (IASI) and PM_{2.5} (in-situ) concentrations on same day. "Independent" - Increase in NH₃ (IASI) concentration not corresponding to an increase in PM_{2.5} (in-situ) concentration on same day. As an example, for the Cologne Munich metropolitan area, the temperature and boundary layer height for the "Simultaneous" case (11.7±6.8 °C and 500.4±166.5 m, respectively) (14.8±8.3 °C and 557.9±193.4 m, respectively) is lower than for the "Independent" case (13.4±6 °C and 628.9±274.3 m, respectively) (15.5±5.4 °C and 599.8±196.3 m, respectively). In addition to low temperature, low boundary layer height results in higher pollutant concentrations and can thus result in more intense atmospheric chemical reactions. We found similar results for other metropolitan areas, but with different absolute values (Fig. 6-(d)) (Table C1). The regional differences are unsurprising, because other factors also influence the formation of PM_{2.5} from NH₃ (e.g., other precursor concentrations such as NO_X and SO_X). However, these findings support previous studies and imply that low temperature and low boundary layer height are most favorable for the formation of PM_{2.5} during the periods of high NH₃. GC also simulates the high spring PM_{2.5} concentrations that have been observed, with high ammonium (NH₄) concentrations (Fig. 6 (c)).

5 Conclusions

Our study estimates the influence of anthropogenic emission reductions on $PM_{2.5}$ concentration changes during the 2020 lock-down period in German metropolitan areas. Mean meteorology accounted for $PM_{2.5}$ concentrations decreased by 5 % during the 2020 lockdown period (spring) compared to the corresponding period in 2019. However, during the 2020 pre-lockdown period (winter), meteorology accounted for $PM_{2.5}$ concentrations are 19 % lower than in 2019. Meanwhile, meteorology accounted for NO_2 levels decreased 23 % during the 2020 lockdown period, which is a larger decrease than 2020 pre-lockdown period compared to 2019 (9 %). No significant change in meteorology accounted for SO_2 and CO concentrations were observed during the 2020 lockdown period, compared to 2019.

The GC model with the COVID-19 emission reduction scenario based on observations (23 % reduction in NO_X emission with unchanged VOC and SO_2) supports our findings of only a marginal decrease in $PM_{2.5}$ and a significant decrease in NO_2 levels. Due to being in a NO_X saturated ozone production regime, the GC lockdown OH and O_3 concentrations increased by 15 % and 9 %, respectively, compared to BAU levels. However, O_X analysis suggest that the only increase in ozone during the day-time is due to increased ozone production efficiency via NO_X saturated regime chemistry, whereas the increase at night-time is due to decreased NO titration. Despite an increase in OH radicals, the GC lockdown PM nitrate formation decreased significantly during the day, due to a significant decrease in NO_2 , compared to the BAU scenario. Increased night-time ozone, however, results in increased night-time NO_3 , despite decreased NO_2 , in turn, resulting in slightly increased night-time N_2O_5 concentration and only a small change in night-time PM nitrate. Overall this results in a small decrease in daily PM nitrate. In addition, the increased OH concentration results in a marginal increase of sulfate formation. Nitrate, sulfate, ammonium and organic aerosol are the major secondary components of $PM_{2.5}$. The decreased day-time PM nitrate is partially offset by the enhanced PM sulfate, and there is no significant impact from slightly decreased PM ammonium and no change in organic aerosol, resulting in a marginal decrease in $PM_{2.5}$ concentrations during the lockdown period.

Based on our findings, we suggest that additional emission control measures aimed at reducing ozone pollution be implemented which should also help reduce PM. A concurrent reduction of NO_X and VOCs emissions should occur. Otherwise, ozone levels will rise as NO_X emissions drop, increasing oxidizing capacity, until a NO_X limited ozone production regime is reached. We also addressed the annual spring $PM_{2.5}$ pollution episodes in German metropolitan areas, which are associated with high NH_3 concentrations. North-West Germany is a hot-spot of NH_3 emissions, primarily emitted from livestock farming and intensive agricultural activities (fertilizer application), with high NH_3 concentrations in the early spring and summer months. Winter and spring meteorological conditions are more favorable for $PM_{2.5}$ formation from NH_3 than summer. Unsurprisingly, low temperature (and low boundary layer height) is shown to be a favorable meteorological condition for the formation of $PM_{2.5}$ from NH_3 . Regulation of NH_3 emissions, primarily from agriculture, has the potential to reduce $PM_{2.5}$ pollution significantly in German metropolitan areas.

410

420

425

In this study, a COVID-19 emission reduction scenario was created using meteorology accounted for proxy pollutant concentration changes, assuming that observed proxy pollutant concentration changes are due to the combined direct effects of emission and meteorology changes. Our GC modeling study work reflects the assumed direct relationship between changes in meteorology accounted for NO_2 concentration and changes in NO_X emission. This work also shows a direct relationship between changes in meteorology accounted for SO_2 (and CO) concentration and changes in SO_X (and CO) emission. However, due to the non-linear feedback system in atmospheric chemistry, this assumption should be investigated further. Because of their similar sources, we use CO concentration as a proxy for anthropogenic VOC concentration. However, this is debatable because VOC is more reactive than CO. We call for further advancements in estimating the emission changes during the lock-down period, which would allow us to estimate the precise sensitivity of $PM_{2.5}$ to changes in emissions from various sources and comparison of VOC emission inventories with observations. This will help in the implementation of appropriate air quality regulation strategies in the future. Organic aerosol accounts for nearly 30 % of total $PM_{2.5}$, which could be influenced by both primary/secondary biogenic and anthropogenic sources. However, our study is limited to examining the effects of NO_X emission changes on $PM_{2.5}$ formation. Therefore, future studies on VOC emission changes on OA formation during high PM pollution episodes, particularly in the spring, will be more important in mitigating PM pollution.

430 Data availability. Hourly measurements of in-situ NO₂, O₃, PM_{2.5}, SO₂ and CO data are downloaded from (https://discomap.eea.europa.eu/map/fme/AirQualityExport.htm). The TROPOMI SO₂ data are obtained from https://s5phub.copernicus.eu/. The IASI NH₃ data are obtained from https://iasi.aeris-data.fr/catalog/. Hourly ERA5 meteorological data are available at https://cds.climate.copernicus.eu/.

Table A1. The statistical evaluation (R, RMSE and mean bias) of the GC model performance ($PM_{2.5}$) for the 2019 study period (January 1 to May 31).

| Metropolitan area | Correlation coefficient (R) | RMSE (μg m ⁻³) | Mean bias | |
|--------------------|-----------------------------|----------------------------|----------------------------|--|
| Wictropolitan area | Correlation coefficient (K) | RWISE (μg III) | (GC – insitu / insitu) (%) | |
| Bremen | 0.6 | 8.7 | -18.9 | |
| Cologne | 0.5 | 11 | 11.7 | |
| Dresden | 0.56 | 9.2 | -18.8 | |
| Dusseldorf | 0.53 | 10.5 | -15.7 | |
| Frankfurt | 0.58 | 9.3 | -37.4 | |
| Hamburg | 0.67 | 8 | -12.7 | |
| Hanover | 0.59 | 7.9 | -13.1 | |
| Leipzig | 0.39 | 8.4 | -28.6 | |
| Munich | 0.5 | 8.5 | -18.6 | |
| Stuttgart | 0.53 | 8.6 | -16.1 | |

Table B1. The statistical evaluation (R, RMSE and mean bias) of the GC model performance (nitrate and ammonium in $PM_{2.5}$) for the 2019 study period (January 1 to May 31). For this comparison, data from the urban measurement station (14.33°E, 51.75°N) in Germany is used.

| Species | Correlation coefficient | RMSE | Mean bias | |
|----------|-------------------------|-------------------------|----------------------------|--|
| species | (R) | $(\mu \text{g m}^{-3})$ | (GC – insitu / insitu) (%) | |
| Nitrate | 0.51 | 2.33 | -32.1 | |
| Ammonium | 0.45 | 1.34 | 37 | |

Table C1. The Statistical distribution of meteorological parameters for the cases "Independent" (each row top) and "Simultaneous" (each row bottom) in ten German metropolitan areas for 2018 and 2019.

| Metropolitan area | Number of days | Wind speed | Temperature | RH | PBL height |
|-------------------|----------------|---------------|----------------|-----------------|-------------------|
| Wetropontan area | rumber of days | (m/s) | (° C) | (%) | (m) |
| | 17 | 4.3 ± 2.1 | 13.6 ± 5.8 | 62.3 ± 14.1 | 625.5 ± 211.1 |
| Bremen | | | | | |
| | 27 | 4.5 ± 2 | 11.5 ± 7 | 67.3 ± 16 | 541 ± 212.5 |
| | 16 | 3±2.2 | 13.4±6.1 | 74.3±11.4 | 628.9±274.31 |
| Cologne | | | | | |
| | 24 | 3.2±1.7 | 11.7±6.8 | 65.3±14.4 | 500.4±166.4 |
| | 24 | 1.9±1.1 | 14.9±6.9 | 68.6 ± 12.8 | 578.9±220.7 |
| Dresden | | | | | |
| | 20 | 2.4±0.8 | 11.1±7.4 | 66.3±11 | 592.1±208.8 |
| | 10 | 3.4±2.1 | 13.2±4.8 | 69±11.3 | 732.1±311.8 |
| Dusseldorf | | | | | |
| | 30 | 3.4±1.8 | 13.5±5.6 | 66.2±13.5 | 494±168 |
| | 18 | 3.2±1.8 | 13.1±6.3 | 64.9 ± 13.2 | 695.2±284.1 |
| Frankfurt | | | | | |
| | 21 | 2.2±1.1 | 13.1±6.6 | 63.6±13.6 | 442.8±194.5 |
| | 14 | 5.4±2.5 | 13.7±6.5 | 57.5 ± 11.8 | 705.3±249.2 |
| Hamburg | | | | | |
| | 27 | 5.2±2.3 | 11.1±3.3 | 67.7±15 | 674.1±262 |
| | 14 | 3.2±2 | 14.2±7.8 | 62.5 ± 10.4 | 697.5±210.2 |
| Hannover | | | | | |
| | 24 | 3.8±1.9 | 9.3±7.6 | 67.6±13.1 | 557.5±176.3 |
| | 18 | 2.9 ± 1.4 | 14.9±8 | 63.7 ± 12.7 | 674.6±206.3 |
| Leipzig | | | | | |
| | 30 | 3.4±1.6 | 11.2±7.1 | 61.9 ± 10.8 | 532.3±227.3 |
| | 26 | 2±1.1 | 15.5±5.4 | 71.5 ± 12.3 | 599.8±196.3 |
| Munich | | | | | |
| | 17 | 1.6±0.8 | 14.8±8.3 | 65.4±9.8 | 557.9±193.4 |
| | 22 | 1.9±0.9 | 13.8±6.4 | 71.7±11 | 600.7±234.9 |
| Stuttgart | | | | | |
| | 22 | 1.5±0.6 | 13.7±6.3 | 67.3±12.9 | 449±191.1 |

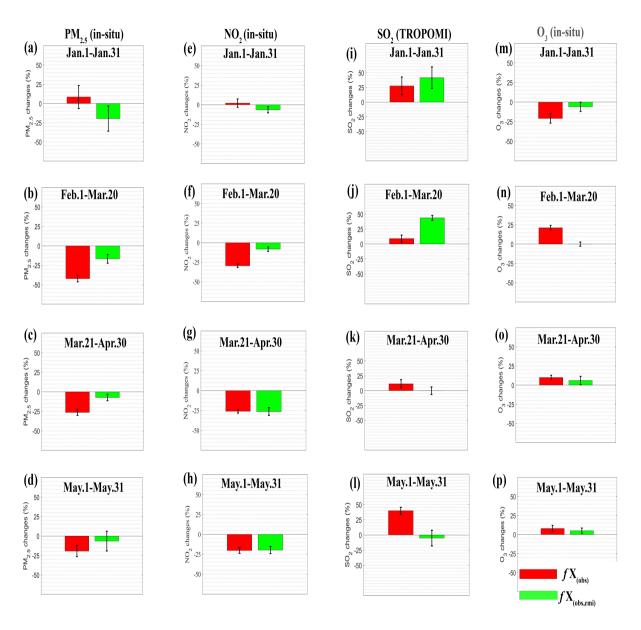


Figure A1. Meteorology unaccounted for (red) and meteorology accounted for (green) mean changes in $PM_{2.5}$, NO_2 , SO_2 and O_3 concentrations between 2020 and 2019 in ten German metropolitan areas. Error bars represent the 1 σ of mean of ten metropolitan areas.

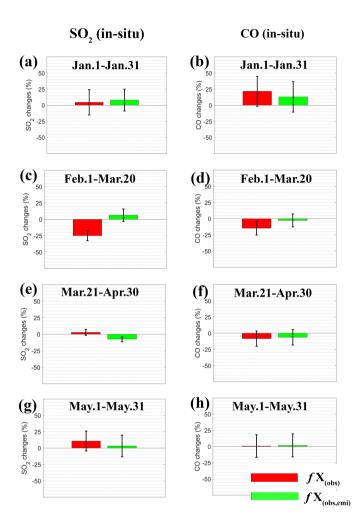


Figure B1. Meteorology unaccounted for (red) and meteorology accounted for (green) mean changes in in-situ SO₂ (Bremen, Dresden, Frankfurt, Hamburg and Leipzig) and in in-situ CO (Bremen, Frankfurt, Hamburg, Hanover, Munich and Stuttgart) between 2020 and 2019. Error bars represent the 1 σ of mean of above mentioned metropolitan areas.

Lockdown period (Mar.21-May.31)

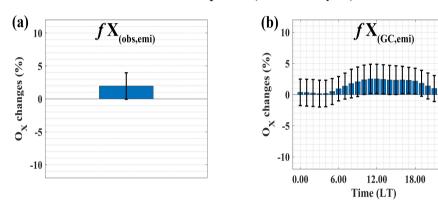


Figure C1. Meteorology accounted for mean changes in in-situ O_X between 2020 and 2019 (left). Diurnal cycle of emission accounted for GC O_X concentration changes between 2020 lockdown and 2020 BAU (no lockdown) scenario (right). Error bars represent the 1 σ of mean of ten metropolitan areas.

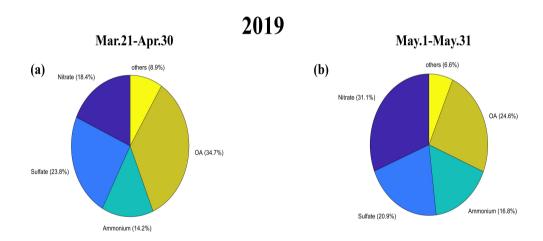


Figure D1. Mean relative contributions of PM_{2.5} species simulated by GC for 2019.

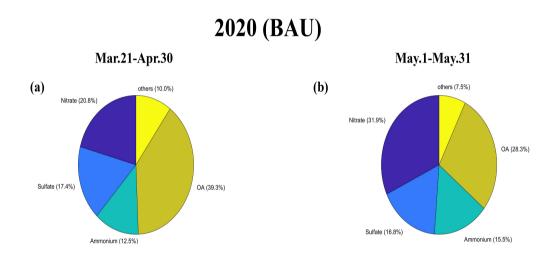


Figure E1. Mean relative contributions of PM_{2.5} species simulated by GC for 2020 (no lockdown).

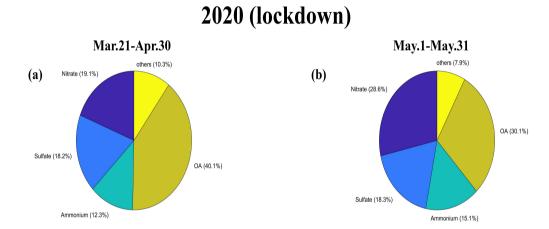


Figure F1. Mean relative contributions of PM_{2.5} species simulated by GC for 2020 (lockdown).

Author contributions. VB, and XB obtained the measurement data; ZQ performed the modeling work; VB analyzed the data and wrote the manuscript draft; JC, and FK supervised the work and edited the manuscript.

435 Competing interests. The authors declare that they have no conflict of interest.

Acknowledgements. Vigneshkumar Balamurugan, Jia Chen, and Xiao Bi are supported by Institute for Advanced Study, Technical University of Munich, through the German Excellence Initiative and the European Union Seventh Framework Program (Grant: 291763) and in part by the German Research Foundation (DFG) (Grant: 419317138). Frank N. Keutsch is funded and supported by the Harvard' Solar Geoengineering Research Program.

440 References

- EUROSAT, Livestock density by NUTS 2 regions, EU-28, 2013. https://ec.europa.eu/eurostat/statistics-explained/index.php?title=File: Livestock_density_by_NUTS_2 regions, EU-28, 2013.png, 2013.
- ESA, Mapping Germany's agricultural landscape. https://www.esa.int/ESA_Multimedia/Images/2017/08/Mapping_Germany_s_agricultural_landscape, 2017.
- 445 SO2 emission (EEA), Indicator Assessment:Sulphur dioxide (SO2) emissions. https://www.eea.europa.eu/data-and-maps/indicators/eea-32-sulphur-dioxide-so2-emissions-1/assessment-3, 2021.
 - Allen, H. M., Draper, D. C., Ayres, B. R., Ault, A., Bondy, A., Takahama, S., Modini, R. L., Baumann, K., Edgerton, E., Knote, C., et al.: Influence of crustal dust and sea spray supermicron particle concentrations and acidity on inorganic NO 3- aerosol during the 2013 Southern Oxidant and Aerosol Study, Atmospheric Chemistry and Physics, 15, 10669–10685, 2015.
- Ansari, A. S. and Pandis, S. N.: Response of inorganic PM to precursor concentrations, Environmental Science & Technology, 32, 2706–2714, 1998.
 - Ayres, B., Allen, H., Draper, D., Brown, S., Wild, R., Jimenez, J., Day, D., Campuzano-Jost, P., Hu, W. d., Gouw, J. d., et al.: Organic nitrate aerosol formation via NO 3+ biogenic volatile organic compounds in the southeastern United States, Atmospheric Chemistry and Physics, 15, 13 377–13 392, 2015.
- Baker, A. K., Beyersdorf, A. J., Doezema, L. A., Katzenstein, A., Meinardi, S., Simpson, I. J., Blake, D. R., and Rowland, F. S.: Measurements of nonmethane hydrocarbons in 28 United States cities, Atmospheric Environment, 42, 170–182, 2008.
 - Balamurugan, V., Chen, J., Qu, Z., Bi, X., Gensheimer, J., Shekhar, A., Bhattacharjee, S., and Keutsch, F. N.: Tropospheric NO2 and O3 response to COVID-19 lockdown restrictions at the national and urban scales in Germany, Journal of Geophysical Research: Atmospheres, 126, 2021.
- 460 Banzhaf, S., Schaap, M., Wichink Kruit, R. J., Denier Van Der Gon, H., Stern, R., and Builtjes, P.: Impact of emission changes on secondary inorganic aerosol episodes across Germany, Atmospheric Chemistry and Physics, 13, 11 675–11 693, 2013.
 - Barré, J., Petetin, H., Colette, A., Guevara, M., Peuch, V.-H., Rouil, L., Engelen, R., Inness, A., Flemming, J., Pérez García-Pando, C., et al.: Estimating lockdown induced European NO 2 changes, Atmospheric Chemistry and Physics Discussions, pp. 1–28, 2020.
 - Bauer, S., Koch, D., Unger, N., Metzger, S., Shindell, D., and Streets, D.: Nitrate aerosols today and in 2030: a global simulation including aerosols and tropospheric ozone, Atmospheric Chemistry and Physics, 7, 5043–5059, 2007.
 - Bauwens, M., Compernolle, S., Stavrakou, T., Müller, J.-F., Van Gent, J., Eskes, H., Levelt, P. F., van der A, R., Veefkind, J., Vlietinck, J., et al.: Impact of coronavirus outbreak on NO2 pollution assessed using TROPOMI and OMI observations, Geophysical Research Letters, 47, e2020GL087 978, 2020.
- Behera, S. N. and Sharma, M.: Investigating the potential role of ammonia in ion chemistry of fine particulate matter formation for an urban environment, Science of the Total Environment, 408, 3569–3575, 2010.
 - Biswal, A., Singh, T., Singh, V., Ravindra, K., and Mor, S.: COVID-19 lockdown and its impact on tropospheric NO2 concentrations over India using satellite-based data, Heliyon, 6, e04 764, 2020.
 - Blanchard, C. L. and Tanenbaum, S. J.: Differences between weekday and weekend air pollutant levels in southern California, Journal of the Air & Waste Management Association, 53, 816–828, 2003.
- 475 Campbell, P. C., Tong, D., Tang, Y., Baker, B., Lee, P., Saylor, R., Stein, A., Ma, S., Lamsal, L., and Qu, Z.: Impacts of the COVID-19 Economic Slowdown on Ozone Pollution in the US, Atmospheric Environment, p. 118713, 2021.

- Carlton, A., Wiedinmyer, C., and Kroll, J.: A review of Secondary Organic Aerosol (SOA) formation from isoprene, Atmospheric Chemistry and Physics, 9, 4987–5005, 2009.
- Cesari, D., De Benedetto, G., Bonasoni, P., Busetto, M., Dinoi, A., Merico, E., Chirizzi, D., Cristofanelli, P., Donateo, A., Grasso, F., et al.:

 Seasonal variability of PM2. 5 and PM10 composition and sources in an urban background site in Southern Italy, Science of the Total Environment, 612, 202–213, 2018.
 - Chan, Y.-C., Evans, M. J., He, P., Holmes, C. D., Jaeglé, L., Kasibhatla, P., Liu, X.-Y., Sherwen, T., Thornton, J. A., Wang, X., et al.: Heterogeneous nitrate production mechanisms in intense haze events in the North China Plain, Journal of Geophysical Research: Atmospheres, 126, e2021JD034688, 2021.
- 485 Chang, W. L., Bhave, P. V., Brown, S. S., Riemer, N., Stutz, J., and Dabdub, D.: Heterogeneous atmospheric chemistry, ambient measurements, and model calculations of N2O5: A review, Aerosol Science and Technology, 45, 665–695, 2011.
 - Chang, W. L., Brown, S. S., Stutz, J., Middlebrook, A. M., Bahreini, R., Wagner, N. L., Dubé, W. P., Pollack, I. B., Ryerson, T. B., and Riemer, N.: Evaluating N2O5 heterogeneous hydrolysis parameterizations for CalNex 2010, Journal of Geophysical Research: Atmospheres, 121, 5051–5070, 2016.
- 490 Clark, H., Bennouna, Y., Tsivlidou, M., Wolff, P., Sauvage, B., Barret, B., Le Flochmoën, E., Blot, R., Boulanger, D., Cousin, J.-M., et al.: The Effects of the COVID-19 Lockdowns on the Composition of the Troposphere as Seen by IAGOS, Atmospheric Chemistry and Physics Discussions, pp. 1–33, 2021.

- Clerbaux, C., Boynard, A., Clarisse, L., George, M., Hadji-Lazaro, J., Herbin, H., Hurtmans, D., Pommier, M., Razavi, A., Turquety, S., et al.: Monitoring of atmospheric composition using the thermal infrared IASI/MetOp sounder, Atmospheric Chemistry and Physics, 9, 6041–6054, 2009.
- Collivignarelli, M. C., Abbà, A., Bertanza, G., Pedrazzani, R., Ricciardi, P., and Miino, M. C.: Lockdown for CoViD-2019 in Milan: What are the effects on air quality?, Science of the total environment, 732, 139 280, 2020.
- Deroubaix, A., Brasseur, G., Gaubert, B., Labuhn, I., Menut, L., Siour, G., and Tuccella, P.: Response of surface ozone concentration to emission reduction and meteorology during the COVID-19 lockdown in Europe, Meteorological Applications, 28, e1990, 2021.
- 500 Dietrich, F., Chen, J., Voggenreiter, B., Aigner, P., Nachtigall, N., and Reger, B.: MUCCnet: Munich Urban Carbon Column network, Atmospheric Measurement Techniques, 14, 1111–1126, 2021.
 - Doumbia, T., Granier, C., Elguindi, N., Bouarar, I., Darras, S., Brasseur, G., Gaubert, B., Liu, Y., Shi, X., Stavrakou, T., et al.: Changes in global air pollutant emissions during the COVID-19 pandemic: a dataset for atmospheric chemistry modeling, Earth System Science Data Discussions, pp. 1–26, 2021.
- Erisman, J. and Schaap, M.: The need for ammonia abatement with respect to secondary PM reductions in Europe, Environmental Pollution, 129, 159–163, 2004.
 - Ernst, J. and Massey, H.: The effects of several factors on volatilization of ammonia formed from urea in the soil, Soil Science Society of America Journal, 24, 87–90, 1960.
- Field, R. D., Hickman, J. E., Geogdzhayev, I. V., Tsigaridis, K., and Bauer, S. E.: Changes in satellite retrievals of atmospheric composition over eastern China during the 2020 COVID-19 lockdowns, Atmospheric Chemistry and Physics Discussions, pp. 1–31, 2020.
 - Filonchyk, M., Hurynovich, V., and Yan, H.: Impact of Covid-19 lockdown on air quality in the Poland, Eastern Europe, Environmental Research, 198, 110 454, 2021.
 - Fisher, J. A., Jacob, D. J., Travis, K. R., Kim, P. S., Marais, E. A., Chan Miller, C., Yu, K., Zhu, L., Yantosca, R. M., Sulprizio, M. P., et al.: Organic nitrate chemistry and its implications for nitrogen budgets in an isoprene-and monoterpene-rich atmosphere: constraints from

- aircraft (SEAC 4 RS) and ground-based (SOAS) observations in the Southeast US, Atmospheric chemistry and physics, 16, 5969–5991, 2016.
 - Fortems-Cheiney, A., Dufour, G., Hamaoui-Laguel, L., Foret, G., Siour, G., Van Damme, M., Meleux, F., Coheur, P.-F., Clerbaux, C., Clarisse, L., et al.: Unaccounted variability in NH3 agricultural sources detected by IASI contributing to European spring haze episode, Geophysical Research Letters, 43, 5475–5482, 2016.
- 520 Fujita, E. M., Campbell, D. E., Zielinska, B., Sagebiel, J. C., Bowen, J. L., Goliff, W. S., Stockwell, W. R., and Lawson, D. R.: Diurnal and weekday variations in the source contributions of ozone precursors in California's South Coast Air Basin, Journal of the Air & Waste Management Association, 53, 844–863, 2003.

535

- Fuzzi, S., Baltensperger, U., Carslaw, K., Decesari, S., Denier van der Gon, H., Facchini, M. C., Fowler, D., Koren, I., Langford, B., Lohmann, U., et al.: Particulate matter, air quality and climate: lessons learned and future needs, Atmospheric chemistry and physics, 15, 8217–8299, 2015.
- Gaubert, B., Bouarar, I., Doumbia, T., Liu, Y., Stavrakou, T., Deroubaix, A., Darras, S., Elguindi, N., Granier, C., Lacey, F., et al.: Global changes in secondary atmospheric pollutants during the 2020 COVID-19 pandemic, Journal of Geophysical Research: Atmospheres, 126, e2020JD034 213, 2021.
- Gensheimer, J., Turner, A. J., Shekhar, A., Wenzel, A., Keutsch, F. N., and Chen, J.: What are different measures of mobility telling us about surface transportation CO2 emissions during the COVID-19 pandemic?, Journal of Geophysical Research: Atmospheres, p. e2021JD034664, 2021.
 - Geyer, A., Alicke, B., Konrad, S., Schmitz, T., Stutz, J., and Platt, U.: Chemistry and oxidation capacity of the nitrate radical in the continental boundary layer near Berlin, Journal of Geophysical Research: Atmospheres, 106, 8013–8025, 2001.
 - Goldberg, D. L., Anenberg, S. C., Griffin, D., McLinden, C. A., Lu, Z., and Streets, D. G.: Disentangling the impact of the COVID-19 lockdowns on urban NO2 from natural variability, Geophysical Research Letters, 47, e2020GL089 269, 2020.
 - Grange, S. K., Lee, J. D., Drysdale, W. S., Lewis, A. C., Hueglin, C., Emmenegger, L., and Carslaw, D. C.: COVID-19 lockdowns highlight a risk of increasing ozone pollution in European urban areas, Atmospheric Chemistry and Physics Discussions, pp. 1–25, 2020.
 - Guevara, M., Jorba, O., Soret, A., Petetin, H., Bowdalo, D., Serradell, K., Tena, C., Denier van der Gon, H., Kuenen, J., Peuch, V.-H., et al.: Time-resolved emission reductions for atmospheric chemistry modelling in Europe during the COVID-19 lockdowns, Atmospheric Chemistry and Physics, 21, 773–797, 2021.
 - Hallquist, M., Wenger, J. C., Baltensperger, U., Rudich, Y., Simpson, D., Claeys, M., Dommen, J., Donahue, N., George, C., Goldstein, A., et al.: The formation, properties and impact of secondary organic aerosol: current and emerging issues, Atmospheric chemistry and physics, 9, 5155–5236, 2009.
- Hammer, M. S., van Donkelaar, A., Martin, R. V., McDuffie, E. E., Lyapustin, A., Sayer, A. M., Hsu, N. C., Levy, R. C., Garay, M. J.,
 Kalashnikova, O. V., et al.: Effects of COVID-19 lockdowns on fine particulate matter concentrations, Science Advances, 7, eabg7670,
 2021.
 - He, C., Hong, S., Zhang, L., Mu, H., Xin, A., Zhou, Y., Liu, J., Liu, N., Su, Y., Tian, Y., et al.: Global, continental, and national variation in PM2. 5, O3, and NO2 concentrations during the early 2020 COVID-19 lockdown, Atmospheric pollution research, 12, 136–145, 2021.
- Hersbach, H., Bell, B., Berrisford, P., Hirahara, S., Horányi, A., Muñoz-Sabater, J., Nicolas, J., Peubey, C., Radu, R., Schepers, D., et al.:

 The ERA5 global reanalysis, Quarterly Journal of the Royal Meteorological Society, 146, 1999–2049, 2020.
 - Higham, J., Ramírez, C. A., Green, M., and Morse, A.: UK COVID-19 lockdown: 100 days of air pollution reduction?, Air Quality, Atmosphere & Health, 14, 325–332, 2021.

- Hoesly, R. M., Smith, S. J., Feng, L., Klimont, Z., Janssens-Maenhout, G., Pitkanen, T., Seibert, J. J., Vu, L., Andres, R. J., Bolt, R. M., et al.:

 Historical (1750–2014) anthropogenic emissions of reactive gases and aerosols from the Community Emissions Data System (CEDS),

 Geoscientific Model Development, 11, 369–408, 2018.
 - Hörmann, S., Jammoul, F., Kuenzer, T., and Stadlober, E.: Separating the impact of gradual lockdown measures on air pollutants from seasonal variability, Atmospheric pollution research, 12, 84–92, 2021.
 - Huang, X., Ding, A., Gao, J., Zheng, B., Zhou, D., Qi, X., Tang, R., Wang, J., Ren, C., Nie, W., et al.: Enhanced secondary pollution offset reduction of primary emissions during COVID-19 lockdown in China, National Science Review, 8, nwaa137, 2021.
- Hudman, R., Moore, N., Martin, R., Russell, A., Mebust, A., Valin, L., and Cohen, R.: A mechanistic model of global soil nitric oxide emissions: implementation and space based-constraints., Atmospheric Chemistry & Physics Discussions, 12, 2012.
 - Jacob, D. J.: Introduction to atmospheric chemistry, Princeton University Press, 1999.
 - Jacobson, M. Z.: Fundamentals of atmospheric modeling, Cambridge university press, 1999.
- Jiménez, P., Parra, R., Gasso, S., and Baldasano, J. M.: Modeling the ozone weekend effect in very complex terrains: a case study in the

 Northeastern Iberian Peninsula, Atmospheric environment, 39, 429–444, 2005.
 - Juda-Rezler, K., Reizer, M., Maciejewska, K., Błaszczak, B., and Klejnowski, K.: Characterization of atmospheric PM2. 5 sources at a Central European urban background site, Science of the Total Environment, 713, 136729, 2020.
 - Kang, M., Zhang, J., Zhang, H., and Ying, Q.: On the Relevancy of Observed Ozone Increase during COVID-19 Lockdown to Summertime Ozone and PM2. 5 Control Policies in China. Environmental Science & Technology Letters. 8, 289–294, 2021.
- Kang, Y.-H., You, S., Bae, M., Kim, E., Son, K., Bae, C., Kim, Y., Kim, B.-U., Kim, H. C., and Kim, S.: The impacts of COVID-19, meteorology, and emission control policies on PM 2.5 drops in Northeast Asia, Scientific reports, 10, 1–8, 2020.
 - Keller, C. A., Evans, M. J., Knowland, K. E., Hasenkopf, C. A., Modekurty, S., Lucchesi, R. A., Oda, T., Franca, B. B., Mandarino, F. C., Díaz Suárez, M. V., et al.: Global impact of COVID-19 restrictions on the surface concentrations of nitrogen dioxide and ozone, Atmospheric Chemistry and Physics, 21, 3555–3592, 2021.
- Kim, Y. J., Spak, S. N., Carmichael, G. R., Riemer, N., and Stanier, C. O.: Modeled aerosol nitrate formation pathways during wintertime in the Great Lakes region of North America, Journal of Geophysical Research: Atmospheres, 119, 12–420, 2014.
 - Koukouli, M.-E., Skoulidou, I., Karavias, A., Parcharidis, I., Balis, D., Manders, A., Segers, A., Eskes, H., and van Geffen, J.: Sudden changes in nitrogen dioxide emissions over Greece due to lockdown after the outbreak of COVID-19, Atmospheric Chemistry and Physics, 21, 1759–1774, 2021.
- 580 Kroll, J. H., Heald, C. L., Cappa, C. D., Farmer, D. K., Fry, J. L., Murphy, J. G., and Steiner, A. L.: The complex chemical effects of COVID-19 shutdowns on air quality, Nature Chemistry, 12, 777–779, 2020.
 - Kuttippurath, J., Singh, A., Dash, S., Mallick, N., Clerbaux, C., Van Damme, M., Clarisse, L., Coheur, P.-F., Raj, S., Abbhishek, K., et al.: Record high levels of atmospheric ammonia over India: Spatial and temporal analyses, Science of the Total Environment, 740, 139 986, 2020.
- Le Quéré, C., Jackson, R. B., Jones, M. W., Smith, A. J., Abernethy, S., Andrew, R. M., De-Gol, A. J., Willis, D. R., Shan, Y., Canadell, J. G., et al.: Temporary reduction in daily global CO 2 emissions during the COVID-19 forced confinement, Nature Climate Change, pp. 1–7, 2020.
 - Lee, J. D., Drysdale, W. S., Finch, D. P., Wilde, S. E., and Palmer, P. I.: UK surface NO 2 levels dropped by 42% during the COVID-19 lockdown: impact on surface O 3, Atmospheric Chemistry and Physics, 20, 15743–15759, 2020.

- Liu, L., Bei, N., Hu, B., Wu, J., Liu, S., Li, X., Wang, R., Liu, Z., Shen, Z., and Li, G.: Wintertime nitrate formation pathways in the north China plain: Importance of N2O5 heterogeneous hydrolysis, Environmental Pollution, 266, 115 287, 2020.
 - Marais, E. A., Jacob, D. J., Jimenez, J. L., Campuzano-Jost, P., Day, D. A., Hu, W., Krechmer, J., Zhu, L., Kim, P. S., Miller, C. C., et al.: Aqueous-phase mechanism for secondary organic aerosol formation from isoprene: application to the southeast United States and cobenefit of SO 2 emission controls, Atmospheric Chemistry and Physics, 16, 1603–1618, 2016.
- Matthias, V., Quante, M., Arndt, J. A., Badeke, R., Fink, L., Petrik, R., Feldner, J., Schwarzkopf, D., Link, E.-M., Ramacher, M. O., et al.: The role of emission reductions and the meteorological situation for air quality improvements during the COVID-19 lockdown period in Central Europe, Atmospheric Chemistry and Physics, 21, 13 931–13 971, 2021.
 - Mendez-Espinosa, J. F., Rojas, N. Y., Vargas, J., Pachón, J. E., Belalcazar, L. C., and Ramírez, O.: Air quality variations in Northern South America during the COVID-19 lockdown, Science of the Total Environment, 749, 141 621, 2020.
- 600 Menut, L., Bessagnet, B., Siour, G., Mailler, S., Pennel, R., and Cholakian, A.: Impact of lockdown measures to combat Covid-19 on air quality over western Europe, Science of the Total Environment, 741, 140 426, 2020.
 - Mollner, A. K., Valluvadasan, S., Feng, L., Sprague, M. K., Okumura, M., Milligan, D. B., Bloss, W. J., Sander, S. P., Martien, P. T., Harley, R. A., et al.: Rate of gas phase association of hydroxyl radical and nitrogen dioxide, Science, 330, 646–649, 2010.
 - Mozurkewich, M.: The dissociation constant of ammonium nitrate and its dependence on temperature, relative humidity and particle size, Atmospheric Environment. Part A. General Topics, 27, 261–270, 1993.
 - Murray, L. T.: Lightning NO x and impacts on air quality, Current Pollution Reports, 2, 115–133, 2016.

- Ordóñez, C., Garrido-Perez, J. M., and García-Herrera, R.: Early spring near-surface ozone in Europe during the COVID-19 shutdown: Meteorological effects outweigh emission changes, Science of the total environment, 747, 141 322, 2020.
- Pathakoti, M., Muppalla, A., Hazra, S., Dangeti, M., Shekhar, R., Jella, S., Mullapudi, S. S., Andugulapati, P., and Vijayasundaram, U.: An assessment of the impact of a nation-wide lockdown on air pollution—a remote sensing perspective over India, Atmospheric Chemistry and Physics Discussions, pp. 1–16, 2020.
 - Pay, M. T., Jiménez-Guerrero, P., and Baldasano, J. M.: Assessing sensitivity regimes of secondary inorganic aerosol formation in Europe with the CALIOPE-EU modeling system, Atmospheric Environment, 51, 146–164, 2012.
- Petetin, H., Sciare, J., Bressi, M., Gros, V., Rosso, A., Sanchez, O., Sarda-Estève, R., Petit, J.-E., and Beekmann, M.: Assessing the ammonium nitrate formation regime in the Paris megacity and its representation in the CHIMERE model, Atmospheric Chemistry and Physics, 16, 10419–10440, 2016.
 - Petetin, H., Bowdalo, D., Soret, A., Guevara, M., Jorba, O., Serradell, K., and Pérez García-Pando, C.: Meteorology-normalized impact of the COVID-19 lockdown upon NO 2 pollution in Spain, Atmospheric Chemistry and Physics, 20, 11 119–11 141, 2020.
- Putaud, J.-P., Pozzoli, L., Pisoni, E., Martins Dos Santos, S., Lagler, F., Lanzani, G., Dal Santo, U., and Colette, A.: Impacts of the COVID-19 lockdown on air pollution at regional and urban background sites in northern Italy, Atmospheric Chemistry and Physics, 21, 7597–7609, 2021.
 - Qu, Z., Jacob, D. J., Silvern, R. F., Shah, V., Campbell, P. C., Valin, L. C., and Murray, L. T.: US COVID-19 shutdown demonstrates importance of background NO2 in inferring NOx emissions from satellite NO2 observations, Geophysical research letters, 48, e2021GL092783, 2021.
- Ramanantenasoa, M. M. J., Gilliot, J.-M., Mignolet, C., Bedos, C., Mathias, E., Eglin, T., Makowski, D., and Génermont, S.: A new framework to estimate spatio-temporal ammonia emissions due to nitrogen fertilization in France, Science of the Total Environment, 645, 205–219, 2018.

- Renner, E. and Wolke, R.: Modelling the formation and atmospheric transport of secondary inorganic aerosols with special attention to regions with high ammonia emissions, Atmospheric Environment, 44, 1904–1912, 2010.
- Russell, A. G., Cass, G. R., and Seinfeld, J. H.: On some aspects of nighttime atmospheric chemistry, Environmental science & technology, 20, 1167–1172, 1986.
 - Salameh, D., Detournay, A., Pey, J., Pérez, N., Liguori, F., Saraga, D., Bove, M. C., Brotto, P., Cassola, F., Massabò, D., et al.: PM2. 5 chemical composition in five European Mediterranean cities: a 1-year study, Atmospheric Research, 155, 102–117, 2015.
- Samek, L., Stegowski, Z., Styszko, K., Furman, L., Zimnoch, M., Skiba, A., Kistler, M., Kasper-Giebl, A., Rozanski, K., and Konduracka,
 E.: Seasonal variations of chemical composition of PM2. 5 fraction in the urban area of Krakow, Poland: PMF source attribution, Air
 Ouality, Atmosphere & Health, 13, 89–96, 2020.
 - Scarlat, N., Fahl, F., Dallemand, J.-F., Monforti, F., and Motola, V.: A spatial analysis of biogas potential from manure in Europe, Renewable and Sustainable Energy Reviews, 94, 915–930, 2018.
- Schaap, M., Loon, M. v., Ten Brink, H., Dentener, F., and Builtjes, P.: Secondary inorganic aerosol simulations for Europe with special attention to nitrate, Atmospheric Chemistry and Physics, 4, 857–874, 2004.
 - Schiferl, L. D., Heald, C. L., Damme, M. V., Clarisse, L., Clerbaux, C., Coheur, P.-F., Nowak, J. B., Neuman, J. A., Herndon, S. C., Roscioli, J. R., et al.: Interannual variability of ammonia concentrations over the United States: sources and implications, Atmospheric Chemistry and Physics, 16, 12 305–12 328, 2016.
- Schumann, U., Poll, I., Teoh, R., Koelle, R., Spinielli, E., Molloy, J., Koudis, G. S., Baumann, R., Bugliaro, L., Stettler, M., et al.: Air traffic and contrail changes over Europe during COVID-19: a model study, Atmospheric Chemistry and Physics, 21, 7429–7450, 2021.
 - Seinfeld, J. H. and Pankow, J. F.: Organic atmospheric particulate material, Annual review of physical chemistry, 54, 121-140, 2003.
 - Shah, V., Jaeglé, L., Thornton, J. A., Lopez-Hilfiker, F. D., Lee, B. H., Schroder, J. C., Campuzano-Jost, P., Jimenez, J. L., Guo, H., Sullivan, A. P., et al.: Chemical feedbacks weaken the wintertime response of particulate sulfate and nitrate to emissions reductions over the eastern United States, Proceedings of the National Academy of Sciences, 115, 8110–8115, 2018.
- Shah, V., Jacob, D. J., Li, K., Silvern, R. F., Zhai, S., Liu, M., Lin, J., and Zhang, Q.: Effect of changing NO x lifetime on the seasonality and long-term trends of satellite-observed tropospheric NO 2 columns over China, Atmospheric Chemistry and Physics, 20, 1483–1495, 2020.
 - Sharma, M., Kishore, S., Tripathi, S., and Behera, S.: Role of atmospheric ammonia in the formation of inorganic secondary particulate matter: a study at Kanpur, India, Journal of Atmospheric Chemistry, 58, 1–17, 2007.
- 655 Sillman, S.: The relation between ozone, NOx and hydrocarbons in urban and polluted rural environments, Atmospheric Environment, 33, 1821–1845, 1999.
 - Sillman, S., Logan, J. A., and Wofsy, S. C.: The sensitivity of ozone to nitrogen oxides and hydrocarbons in regional ozone episodes, Journal of Geophysical Research: Atmospheres, 95, 1837–1851, 1990.
- Solberg, S., Walker, S.-E., Schneider, P., and Guerreiro, C.: Quantifying the Impact of the Covid-19 Lockdown Measures on Nitrogen Dioxide

 Levels throughout Europe, Atmosphere, 12, 131, 2021.
 - Souri, A. H., Chance, K., Bak, J., Nowlan, C. R., González Abad, G., Jung, Y., Wong, D. C., Mao, J., and Liu, X.: Unraveling Pathways of Elevated Ozone Induced by the 2020 Lockdown in Europe by an Observationally Constrained Regional Model: Non-Linear Joint Inversion of NO x and VOC Emissions using TROPOMI, Atmospheric Chemistry and Physics Discussions, pp. 1–45, 2021.
- Squizzato, S., Masiol, M., Brunelli, A., Pistollato, S., Tarabotti, E., Rampazzo, G., and Pavoni, B.: Factors determining the formation of secondary inorganic aerosol: a case study in the Po Valley (Italy), Atmospheric chemistry and physics, 13, 1927–1939, 2013.

- Steinfeld, J. I.: Atmospheric chemistry and physics: from air pollution to climate change, Environment: Science and Policy for Sustainable Development, 40, 26–26, 1998.
- Steinmetz, H., Batzdorfer, V., and Bosnjak, M.: The ZPID lockdown measures dataset for Germany, 2020.

- Stephens, S., Madronich, S., Wu, F., Olson, J., Ramos, R., Retama, A., and Munoz, R.: Weekly patterns of México City's surface concentrations of CO, NO x, PM 10 and O 3 during 1986–2007, Atmospheric Chemistry and Physics, 8, 5313–5325, 2008.
 - Tai, A. P., Mickley, L. J., and Jacob, D. J.: Impact of 2000–2050 climate change on fine particulate matter (PM_{2.5}) air quality inferred from a multi-model analysis of meteorological modes, Atmospheric chemistry and physics, 12, 11 329–11 337, 2012.
 - Theys, N., Smedt, I. D., Yu, H., Danckaert, T., Gent, J. v., Hörmann, C., Wagner, T., Hedelt, P., Bauer, H., Romahn, F., et al.: Sulfur dioxide retrievals from TROPOMI onboard Sentinel-5 Precursor: algorithm theoretical basis, Atmospheric Measurement Techniques, 10, 119–153, 2017.
 - Turner, A. J., Kim, J., Fitzmaurice, H., Newman, C., Worthington, K., Chan, K., Wooldridge, P. J., Köehler, P., Frankenberg, C., and Cohen, R. C.: Observed Impacts of COVID-19 on Urban CO 2 Emissions, Geophysical Research Letters, 47, e2020GL090 037, 2020.
- Viatte, C., Wang, T., Damme, M. V., Dammers, E., Meleux, F., Clarisse, L., Shephard, M. W., Whitburn, S., Coheur, P. F., Cady-Pereira, K. E., et al.: Atmospheric ammonia variability and link with particulate matter formation: a case study over the Paris area, Atmospheric Chemistry and Physics, 20, 577–596, 2020.
 - Viatte, C., Petit, J.-E., Yamanouchi, S., Van Damme, M., Doucerain, C., Germain-Piaulenne, E., Gros, V., Favez, O., Clarisse, L., Coheur, P.-F., et al.: Ammonia and PM2. 5 Air Pollution in Paris during the 2020 COVID Lockdown, Atmosphere, 12, 160, 2021.
 - Von Schneidemesser, E., Monks, P. S., and Plass-Duelmer, C.: Global comparison of VOC and CO observations in urban areas, Atmospheric Environment, 44, 5053–5064, 2010.
- Wang, L., Wang, J., and Fang, C.: Assessing the Impact of Lockdown on Atmospheric Ozone Pollution Amid the First Half of 2020 in Shenyang, China, International Journal of Environmental Research and Public Health, 17, 9004, 2020.
 - Wang, S., Nan, J., Shi, C., Fu, Q., Gao, S., Wang, D., Cui, H., Saiz-Lopez, A., and Zhou, B.: Atmospheric ammonia and its impacts on regional air quality over the megacity of Shanghai, China, Scientific reports, 5, 1–13, 2015.
 - Wang, W., van der A, R., Ding, J., van Weele, M., and Cheng, T.: Spatial and temporal changes of the ozone sensitivity in China based on satellite and ground-based observations, Atmospheric Chemistry and Physics, 21, 7253–7269, 2021.
 - Wang, Y., Zhang, Q., He, K., Zhang, Q., and Chai, L.: Sulfate-nitrate-ammonium aerosols over China: response to 2000–2015 emission changes of sulfur dioxide, nitrogen oxides, and ammonia, Atmospheric Chemistry and Physics, 13, 2635–2652, 2013.
 - Watson, J. G., Chow, J. C., Lurmann, F. W., and Musarra, S. P.: Ammonium nitrate, nitric acid, and ammonia equilibrium in wintertime Phoenix, Arizona, Air & Waste, 44, 405–412, 1994.
- Webb, J., Menzi, H., Pain, B., Misselbrook, T., Dämmgen, U., Hendriks, H., and Döhler, H.: Managing ammonia emissions from livestock production in Europe, Environmental pollution, 135, 399–406, 2005.
 - Werf, G. R., Randerson, J. T., Giglio, L., Leeuwen, T. T. v., Chen, Y., Rogers, B. M., Mu, M., Van Marle, M. J., Morton, D. C., Collatz, G. J., et al.: Global fire emissions estimates during 1997–2016, Earth System Science Data, 9, 697–720, 2017.
- Womack, C., McDuffie, E., Edwards, P., Bares, R., de Gouw, J., Docherty, K., Dubé, W., Fibiger, D., Franchin, A., Gilman, J., et al.: An odd oxygen framework for wintertime ammonium nitrate aerosol pollution in urban areas: NOx and VOC control as mitigation strategies, Geophysical Research Letters, 46, 4971–4979, 2019.
 - Wu, S.-Y., Hu, J.-L., Zhang, Y., and Aneja, V. P.: Modeling atmospheric transport and fate of ammonia in North Carolina—Part II: Effect of ammonia emissions on fine particulate matter formation, Atmospheric Environment, 42, 3437–3451, 2008.

- Wu, Y., Gu, B., Erisman, J. W., Reis, S., Fang, Y., Lu, X., and Zhang, X.: PM2. 5 pollution is substantially affected by ammonia emissions in China, Environmental Pollution, 218, 86–94, 2016.
 - Yan, C., Tham, Y. J., Zha, Q., Wang, X., Xue, L., Dai, J., Wang, Z., and Wang, T.: Fast heterogeneous loss of N2O5 leads to significant night-time NOx removal and nitrate aerosol formation at a coastal background environment of southern China, Science of the Total Environment, 677, 637–647, 2019.
- Yarwood, G., Stoeckenius, T. E., Heiken, J. G., and Dunker, A. M.: Modeling weekday/weekend ozone differences in the Los Angeles region for 1997, Journal of the Air & Waste Management Association, 53, 864–875, 2003.
 - Yin, H., Lu, X., Sun, Y., Li, K., Gao, M., Zheng, B., and Liu, C.: Unprecedented decline in summertime surface ozone over eastern China in 2020 comparably attributable to anthropogenic emission reductions and meteorology, Environmental Research Letters, 2021.
 - Zhai, S., Jacob, D. J., Wang, X., Liu, Z., Wen, T., Shah, V., Li, K., Moch, J. M., Bates, K. H., Song, S., et al.: Control of particulate nitrate air pollution in China, Nature Geoscience, pp. 1–7, 2021.
- 715 Zhang, R., Wang, G., Guo, S., Zamora, M. L., Ying, Q., Lin, Y., Wang, W., Hu, M., and Wang, Y.: Formation of urban fine particulate matter, Chemical reviews, 115, 3803–3855, 2015.