



Impacts of the COVID-19 lockdown on air pollution at regional and urban background sites in northern Italy.

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10 Abstract

The COVID-19 lockdown measures gradually implemented in Lombardy (northern Italy) from 23 February 2020 led to a downturn in several economic sectors with possible impacts on air quality. Several communications claimed in the first weeks of March 2020 that the mitigation in air pollution observed at that time was actually related to these lockdown measures, without considering that seasonal variations in emissions and meteorology also influence air quality. To determine the specific impact of lockdown measures on air quality in northern Italy, we compared observations from the European Commission atmospheric observatory of Ispra (regional background) and from the regional environmental protection agency (ARPA) air monitoring stations in the Milan conurbation (urban background) with expected values for these observations using two different approaches. On the one hand, intensive aerosol variables determined from specific aerosol characterisation observations performed in Ispra were compared to their 3-year averages. On the other hand, measured concentrations of atmospheric pollutants (NO₂, PM₁₀, O₃, NO, SO₂) were compared to expected concentrations derived from the Copernicus Atmosphere Monitoring Service Regional (CAMS) ensemble model forecasts, which did not account for lockdown measures. From these comparisons, we show that NO₂ concentrations decreased as a consequence of the lockdown by -30% and -40% on average at the urban and regional background sites, respectively. Unlike NO₂, PM₁₀ concentrations were not significantly affected by lockdown measures. This could be due to any decreases in PM₁₀ (and PM₁₀ precursors) emissions from traffic being compensated for by increases in emissions from domestic heating and/or from changes in the secondary aerosol formation regime resulting from the lockdown measures. The implementation of the lockdown measures also led to an increase in the highest O₃ concentrations at both the urban and regional background sites resulting from reduced titration of O₃ by NO. The relaxation of the lockdown measures beginning in May resulted in "close to expected" NO₂ concentrations in the urban background, and to significant increases in PM₁₀ in comparison to expected concentrations at both regional and urban background sites.

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1 Introduction

The COVID-19 pandemic is an epidemic of coronavirus disease 2019 (COVID-19), of which the outbreak was first identified in Wuhan, China, in late December 2019. The World Health Organization declared COVID-19 a pandemic on 11 March 2020. The first case of COVID-19 in northern Italy was detected on 20 February 2020 in Codogno, about 60 Km south east of Milan (Figure 1). To reduce the virus spreading, the Italian Government quickly adopted a series of measures, such as the quarantine for 10 municipalities, the cancellation of all main public events and the closure of schools and universities in northern Italy (DL, 23 February 2020). The lockdown started in all of Italy on 9 March 2020 (DPCM, 8 March 2020). All commercial and retail activities were closed on 11 March, except for grocery shops and pharmacies (DPCM, 11 March 2020) and it was forbidden to move outside the place of residence, except for health issues or work. Further lockdown measures were decreed on 22 March 2020 (DPCM, 22 March 2020), including the suspension of all non-essential industrial production activities. The lockdown lasted until 4 May 2020 (DPCM, 26 April 2020), when a gradual relaxation of the measures was decided by the government. The re-opening of manufacturing industries and construction sites was allowed, but schools and universities as well as some commercial activities such as restaurants remained closed. Movements from a region to another were still forbidden, but moving short distances to work and to visit relatives was possible. From 18 May 2020, most commercial businesses could re-open and free movement was granted within regional borders (DL, 16 May 2020). This lockdown provided a unique opportunity to determine how such dramatic measures can eventually influence air quality. This is the focus of this paper.

Lombardy, Piedmont and Emilia-Romagna in northern Italy produce roughly 50% of the national Gross Domestic Product (GDP), with Lombardy alone producing 22% of the national GDP (Istat, 2018 data). This economic dynamism (mainly linked to industrial production and service related activities) is associated with significant pollutant emissions, which together with unfavourable conditions for pollution dispersion (due to low wind speeds and particular orography) cause high pollution levels leading to exceedances of the EU standards for nitrogen dioxide (NO₂), particulate matter (PM₁₀) and ozone (O₃) in northern Italy (EEA, 2019). In this area, the impact of lockdown on economic activities were quite important, as illustrated by data relative to the production of electricity and energy for heating, and to transport related activities (ARPA Lombardia, 2020). Compared to 2019, the Italian thermal electricity production (Figure 2) fell in March (-18%), April (-24%) and May 2020 (-16%). The consumption of natural gas by the industrial sector as reported by the Italian natural gas provider (www.snam.it) also fell by roughly -30% at the end of March in comparison to the beginning of March 2020.

Regarding transport, the Monitoring of Polluting Vehicles project (MOVE-IN) managed by the Lombardy region provided data on the traffic changes derived from its monitoring of 'vehicle km' driven by light duty vehicle and passenger cars (for a small number of vehicles compared to the full fleet circulating in the region though). MOVE-IN data show that the number of 'vehicle km' driven by light duty vehicles remained quite constant till 9 March 2020, then dropped by -75% to reach a minimum between 16 March and 13 of April 2020 before returning to 'usual' (i.e. as before the lockdown period) values after 4 May 2020 (ARPA Lombardia, 2020). For private cars, the number of 'vehicle km' driven also decreased by roughly -70% between the beginning and the end of March, and started increasing again after the 4th of May, but with a slower recovery than for light duty vehicles. The number of requests for driving directions (www.apple.com/covid19/mobility) showed similar variations (Figure 2).

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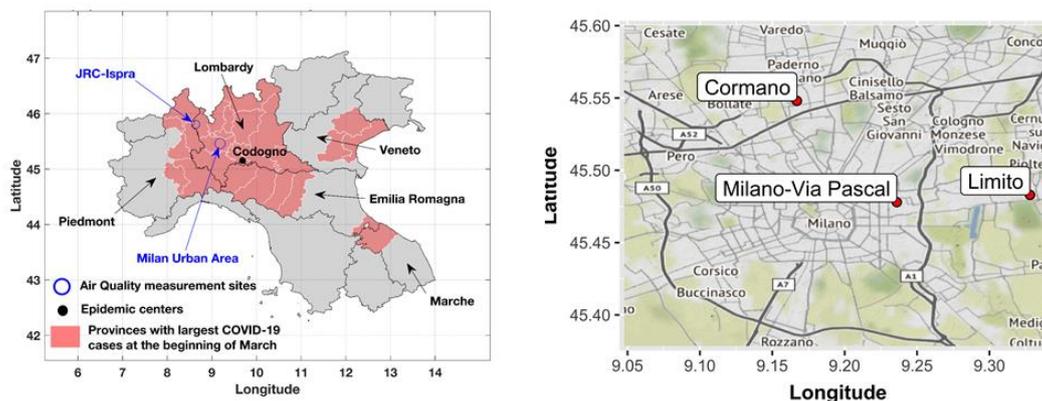


Figure 1: northern Italy areas impacted by the COVID-19 at the beginning of March 2020, and location of the air pollution measurement sites in Ispra and Milan considered in this study. Right hand panel: © [OpenStreetMap contributors](#) 2020. Distributed under a Creative Commons BY-SA License.

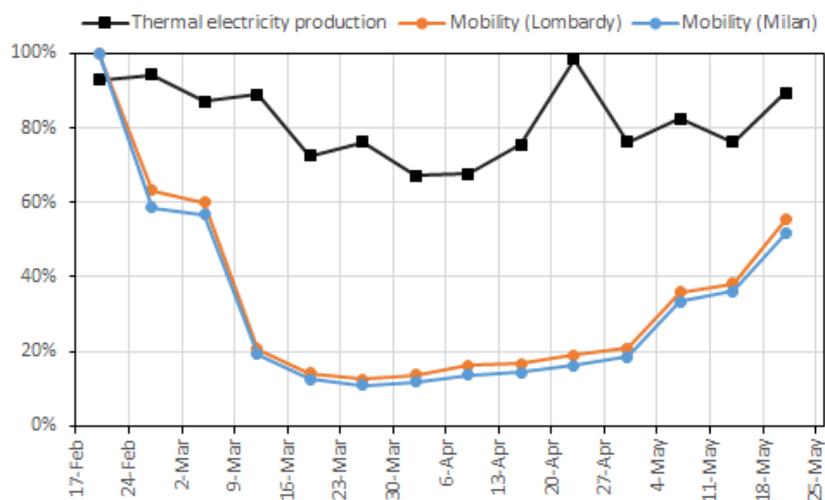
70 Numerous early communications based on preliminary measurement data analyses associated observed improvements in air quality to the lockdown measures taken to contain the spread of the COVID-19 epidemic. In Brazil, the lockdown in Sao Paulo was followed by drastic reductions in NO (up to -77%) and NO₂ (up to -54%) and by an increase in O₃ (approximately + 30%) compared to the previous five-year means for the same period (Nakada et al., 2020). In the Yangtze River Delta Region (China), Li et al. (2020) showed that concentrations of PM_{2.5}, NO₂ and SO₂ decreased by 32%, 45% and 20% during a first lockdown phase, and by 33%, 27% and 7% during a second lockdown phase, compared with the 2017-2019 average for the same period. O₃ also
75 increased in that region. In France, the analysis by INERIS (Institut national de l'environnement industriel et des risques) compared air pollution forecast data (calculated without incorporating changes in emissions due to lockdown measures) with adjusted simulations performed *a posteriori* by assimilating observation data influenced by the lockdown measures. They estimated that NO₂ concentrations were on average approximately 50% lower than expected in France's largest cities (INERIS, 2020).

80 Regarding Italy, maps of NO₂ surface concentrations estimated from satellite data (e.g. Sentinel-5p) were published by several web sites and media showing large reductions of NO₂ concentrations over northern Italy in March 2020 as compared to the previous months and to March 2019 (e.g. Copernicus Atmosphere Monitoring Service headlines published on 17 and 26 of March 2020). Observations and models were also combined in the analysis from the German Aerospace Centre (DLR) which estimated a decrease of about 40% in the total column-integrated NO₂ tropospheric concentrations over northern Italy due to the lockdown
85 measures, using Sentinel-5p data. They also estimated reductions in ground level NO₂ concentrations of about -20 µg m⁻³ (-45%) by comparing ground base observations from 25 stations in Lombardy to a model simulation with pre-lockdown emission levels (DLR, 05/05/2020). In-situ observations also showed reduced ground level NO₂ concentrations as lockdown measures were implemented. The Environmental Protection Agency ARPA Lombardia showed that March 2020 NO₂ concentrations were below the standard deviation calculated from previous years, indicating a possible signal of reduced emissions from traffic and economic
90 sectors (ARPA Lombardia, 2020). The European Environment Agency (EEA) developed a viewer that tracks NO₂ and particulate matter (PM₁₀ and PM_{2.5}) weekly average concentrations (<https://www.eea.europa.eu/themes/air/air-quality-and-covid19>). It shows that NO₂ concentrations in Milan were at least 24 % lower after the lockdown implementation than during previous weeks, and 21% lower compared to the same period in 2019. Similar trends were found in other cities of northern Italy and European



95 countries where strong measures were taken to contain the epidemic. In contrast, no consistent effect of the lockdown measures
on particulate matter (PM_{2.5} and PM₁₀) could be observed in the main European cities (EEA, 2020).

100 Air pollution did decline in northern Italy from February to May in 2020 as it does every year, mainly due to seasonal variations
in emissions and weather conditions. The strength of certain sources does indeed change during the course of the year, like e.g.
domestic heating, while weather conditions influence pollution concentration in diverse ways: advection and dispersion of
pollutants resulting from horizontal winds, dilution of pollutants throughout the mixed boundary layer resulting from convection,
and pollutant lifetimes resulting from photochemical reactions (sun radiation), wet removal (clouds and rain), etc. It is therefore
not straightforward to disentangle the effects of changing emissions due to lockdown measure implementation from those of
seasonal changes in emissions and variability in meteorological conditions between different seasons and different years. In the
present study we determine how much of the changes in air pollution observed during the lockdown period in northern Italy
were actually due to lockdown measures, independently from expected variations in pollutants' emissions, lifetime and
105 dispersion. Our results are based on comparisons between air pollution observation data from Ispra (regional background site)
and the Milan conurbation (urban background sites) with CAMS-Ensemble model forecast data for the same sites. To help
understand the effect of the lockdown measures in regional background area, we also use 4 years of specific aerosol
measurements from Ispra.



110 *Figure 2: Variations in activities resulting from lockdown measures (2020). Percentages are calculated in comparison with 2019
data for thermal energy production (source: www.terna.it) and in comparison with data from the third week of February 2020 for
mobility data (source: www.apple.com/covid19/mobility).*



2 Material and methods

115 Model and observation air pollution data from 4 sites located in Lombardy covering the time periods 17 February – 24 May
2019 and 2020 were collected and analysed. We selected three sites located in the Milan conurbation as representative of the
Milan urban background, and the site of Ispra as representative of the regional background of the upper Po Valley (Figure 1).
Ground level concentrations of NO, NO₂, SO₂, O₃ and PM₁₀ as measured in-situ at the monitoring stations and as calculated by
the CAMS ensemble model forecast were considered. Particle number size distribution and aerosol light absorption Ångström
120 exponent data from Ispra for the 2017-2020 period were also utilised.

2.1 Site description

The European Commission Atmospheric Observatory (ECA_{tmO}) has been operated in Ispra (45.815 N, 8.636 E, 209 m a.s.l.)
since November 1985. It has contributed to the CLRTAP-EMEP (co-operative programme for monitoring and evaluation of the
long-range transmission of air pollutants in Europe under the Convention on Long-range Transboundary Air Pollution) and WMO-
125 GAW (World Meteorological Organization – Global Atmosphere Watch) air pollution measurement programmes for several
decades and to the European Research Infrastructures ICOS (Integrated Carbon Observation System) and ACTRIS (Research
Infrastructure for the observation of Aerosol, Clouds and Trace Gases) for several years. ECA_{tmO} is located on the northwestern
edge of the Po Valley, 20 - 60 km away from major pollution point sources, but still in a densely populated area (ca 500 / km²)
with significant economic activity (GDP per capita = 29000 €; EUROSTAT, 2017). Wood burning for domestic heating is also an
130 important source of particulate matter during the cold period of the year (Gilardoni et al., 2011). Past measurements of
HCHO/NO₂ ratios compared to the threshold values proposed by Tonnensen and Dennis (2000) suggest that the photochemical
production of O₃ is limited by the availability of volatile organic compounds in February – May in Ispra.

The Milan metropolitan area is the second largest densely populated area in Italy (ca 2300 / km²), with a GDP per capita of
about 54000 € (EUROSTAT, 2017), and about 4100 circulating vehicles/km² (ISTAT, 2018). Three stations in the Milan conurbation
135 were selected as representative of the urban background in Milan city, namely ‘Milan via Pascal’ (45.478 N, 9.236 E, 122 m a.s.l.),
‘Cormano’ (45.548 N, 9.167 E, 155 m a.s.l.), and ‘Limite di Pioltello’ (45.483 N, 9.328 E, 123 m a.s.l.). All three stations are operated
by ARPA Lombardia. We selected only urban background stations because pollutant concentrations at traffic sites are hardly
reproducible by regional air quality models with a horizontal resolution of about 10 km. The station in Milan via Pascal is located
near the university and it is considered as the urban background station of the city, while the other two stations are located in
140 the hinterland, near (< 500 m) two main roadways used by commuters at the northern (Cormano, with about 75000 vehicles/day)
and eastern (Limite, 20000 vehicles/day) entrances of the city. Average population densities are 7500 / km², 4500 / km², and
2800 / km² in Milan, in Cormano, and Limite di Pioltello, respectively (ISTAT, 2018).

2.2 Measurements

At ECA_{tmO} in Ispra, on-line in situ air pollution measurements are performed from appropriate inlets located at 6.5 and
145 9 m above ground level for gaseous and particulate pollutants, respectively. The inlet for reactive gas is made of PTFE (i.d. 2.7
cm). The sample residence time in the inlet tube is ca. 2 s. Each analyser samples from the main inlet through a Nafion dryer. In
2019 – 2020, the measurement programme included CO, NO, NO_x, NO₂, SO₂, O₃, non-methane hydrocarbons (until 6 March 2020)
and NH₃ (since 28 January 2020) as gaseous pollutants. The NO_x (i.e. NO and NO₂), SO₂, and O₃ data reported in this work were
obtained with trace level instruments based on IR (1200 nm) chemiluminescence and Mo converter (ThermoFisher 42iTL), UV



150 (214 nm) fluorescence (ThermoFisher 43i TLE), and UV (254 nm) absorption (ThermoFisher 49C), respectively. These instruments are calibrated every 3 months using zero air and certified gas cylinders (NO and SO₂) or a primary standard ozone generator (O₃). In 2019, annual average concentrations of NO, NO₂, SO₂, O₃ and PM₁₀ were 4, 16, 0.4, 38 and 21 µg m⁻³, respectively. Particulate matter is sampled through metal-made inlets characterized by negligible losses. Each instrument samples isokinetically from the main aerosol inlet through Nafion dryers. In 2019 – 2020, the aerosol on-line in situ measurement programme included PM₁₀
155 mass concentration, particle number concentration and number size distribution, particle light extinction, absorption, scattering and backscattering at several wavelengths. The PM₁₀ mass, particle number and light absorption data reported in this work were obtained with a TEOM-FDMS (ThermoFisher 1405-DF), a Differential Mobility Particle Sizer (Home-made Vienna-type Differential Mobility Analyzer + TSI 3772 Condensation Particle Counter) covering the particle mobility diameter range 10 – 800 nm, and a 7-wavelength Aethalometer (Magee AE31), respectively. The TEOM has been calibrated using a standard filter provided by the
160 manufacturer, while the DMPS and the Aethalometer are operated, maintained and controlled according to ACTRIS guidelines (www.actris.eu). They were both calibrated at the specific ACTRIS central facility (www.actris-ecac.eu) on 3-7 June 2019. Near real time data are available from the JRC data catalogue at data.jrc.ec.europa.eu/collection/abcis.

The three stations in the Milan conurbation are part of the ARPA Lombardia air quality network, compliant with Directive 2008/50/EC requirements in terms of measurement methods, macro and micro localisation, and data coverage. Inlets are located
165 at 2.5 m above ground level for all pollutants. The measurement programmes comprise NO, NO_x, NO₂, SO₂, and O₃ at all three sites. Additional measurements include benzene, toluene, xylenes, PM₁₀, PM_{2.5}, B(a)P and NH₃ in Milan, and CO and PM₁₀ in Limoto. Each gas analyser samples from the main inlet through a Nafion dryer. The NO_x data reported in this work for the Milan conurbation were obtained with trace level instruments based on IR (1200 nm) chemiluminescence and Mo converter (Teledyne API 201E, ThermoFischer 42 in and ThermoFischer 42c in in Milano Pascal, Limoto and Cormano, respectively) and O₃ was
170 measured by UV (254 nm) fluorescence (ThermoFischer 49i at all 3 sites). All measurements are performed according to a specific QA/QC programme. All gas monitors are calibrated every 3 months using zero air and certified gas cylinders (NO) and every 6 months using a primary standard ozone generator for O₃. The PM₁₀ mass concentrations in the Milan conurbation reported in this work were measured using beta absorption analyzers (FAI SWAM DC and 5A models in Milano Pascal and Limoto, respectively). The PM analysers are checked for temperature, pressure, flow rates, leaks, and other operational parameters every 3 months. A
175 periodical comparison with gravimetric samples has been performed once yearly in Milano Pascal, and upon a specific audit programme in Limoto. In 2019, annual average concentrations of NO, NO₂, O₃ were respectively 25, 37, 46 µg m⁻³ in Milano Pascal, 29, 45, 46 µg m⁻³ in Cormano and 26, 34, 44 µg m⁻³ in Limoto. For PM₁₀, 2019 annual averages were 29 µg m⁻³ and 31 µg m⁻³ in Milano Pascal and Limoto, respectively. Data are available on line at www.arpalombardia.it.

2.3 CAMS-Ensemble forecast description

180 The Copernicus Atmospheric Monitoring Service (CAMS) provides, daily, 4-day ahead air quality forecasts for Europe from currently nine different regional air quality models (CHIMERE, DEHM, EMEP, EURAD-IM, GEM-AQ, LOTOS-EUROS, MATCH, MOCAGE, SILAM). Forecasts are performed independently by all the individual regional air quality systems. An ensemble (named ‘CAMS-Ensemble forecast’) is calculated from individual model outputs with a median approach (Marécal et al, 2015). This method provides an optimal estimate (Riccio et al, 2007) which is rather insensitive to outliers and generally yields better
185 estimates than the individual models (Galmarini et al., 2018). The outputs of the different individual models are interpolated on a common regular 0.1°x0.1° latitude x longitude grid (about 10 km x 10 km) over the European domain (25°W-45°E, 30°N-72°N). Hourly pollutant concentrations are calculated for altitudes ranging from the 40m-thick surface layer to 5 km. Each air quality



model is based on different chemical (gas and aerosols) and physical parameterisations, but uses the same meteorological drivers as input (the ECMWF Integrated Forecasting System, IFS) and the same anthropogenic emissions data (Kuenen et al., 2014; Denier van der Gon et al., 2015) based on 2011 emission inventories until June 2019, and on 2016 emission inventories afterwards. As the anthropogenic emissions used by the individual models did not change to account for any lockdown measure, the air quality models continued to forecast pollutants' concentrations as if the COVID-19 epidemic had not occurred in 2020.

2.4 Data analysis.

2.4.1 Pollutant concentrations

To determine the specific impact of lockdown measures on concentrations of air pollutants, we compared daily observations (*Obs*) with daily expected concentrations (*Exp*) for the period 17 February – 24 May 2020, which comprises the 8 lockdown weeks (*D* = 9 March – 3 May 2020), the 3 weeks before the beginning (*A* = 17 February - 8 March 2020) and the 3 weeks after the end of the lockdown period (*P* = 4 – 24 May 2020). NO₂, PM₁₀, NO, O₃ and SO₂ observed and expected concentrations are shown in Figure 3. Expected concentrations were derived from 2020 CAMS –Ensemble forecasts, which account for variations in meteorological conditions and seasonal changes in emission source strengths in a “business as usual” world, i.e. without lockdown measures. However, since data from 2019 show that the agreement between CAMS-Ensemble forecasts (*CAMS*₂₀₁₉) and observations (*Obs*₂₀₁₉) improves from February to May (see Figure S1-S3 in Supplement), *CAMS*₂₀₂₀ were corrected for this seasonality. Thus, 2020 daily expected pollutants concentrations *Exp* were calculated as Eq. (1):

$$Exp = \frac{CAMS_{2020}}{CAMS_{2019}} Obs_{2019} \quad (1)$$

The comparison of observations with these expected concentrations for 2020 has the great advantage of being insensitive to the fact that the emissions inventories used to calculate CAMS-Ensemble forecast data for 2019 and 2020 were different. The disadvantage of this approach is that *Obs* and *Exp* cannot be compared to each other on a daily basis, since *Exp* values are affected by random variations in the *Obs*₂₀₁₉/*CAMS*₂₀₁₉ ratio. Therefore 2020 *Obs* and *Exp* data were compared statistically for the 3 periods A, D, and P. Since changes in pollutant emission rates are expected to result in changes in pollutant concentrations in terms of percentages or ratios, statistical analyses were performed on *Obs* / *Exp* daily ratios. We calculated occurrence frequency distributions of the *Obs* / *Exp* ratio using 8 class bins ranging from <0.25 to >2, all equally wide on a logarithmic scale (except the last one when specifically indicated). Cumulative frequencies of occurrence were also plotted to facilitate comparisons (Figure 4). To detect possible specific impacts of lockdown measures on highest concentrations, specific occurrence frequency distributions were also calculated by selecting the 28 days on which CAMS-Ensemble forecast data were greater than the median during the lockdown period. These days are different for each pollutant and each site. The statistical significance of the differences in *Obs* / *Exp* ratios during the lockdown period in comparison with before and after the lockdown period (i.e. between A and D or P and D) was assessed by applying a t-test assuming unequal variances to the means \bar{A} , \bar{P} , and \bar{D} , defined as Eqs (2):

$$\bar{D} = \text{mean} \left(\log \left(\frac{(Obs/CAMS)_{\text{during lockdown}}}{(Obs/CAMS)_{10 \text{ Mars} - 25 \text{ May } 2019}} \right) \right), \bar{A} = \text{mean} \left(\log \left(\frac{(Obs/CAMS)_{\text{before lockdown}}}{(Obs/CAMS)_{17 \text{ Feb} - 9 \text{ Mars } 2019}} \right) \right), \bar{P} = \text{mean} \left(\log \left(\frac{(Obs/CAMS)_{\text{after lockdown}}}{(Obs/CAMS)_{5 - 25 \text{ May } 2019}} \right) \right) \quad (2)$$

The null hypotheses ($\bar{D} = \bar{A}$, and $\bar{D} = \bar{P}$) were tested at the 95% confidence level, and results were used to determine if differences between \bar{D} and \bar{A} and \bar{D} and \bar{P} were statistically significant.



2.4.2 Intensive aerosol variables

To complement our analyses based on pollutant concentrations, we also looked at two characteristics of the atmospheric aerosol measured at ECAtmO in Ispra. The first one is the percentage in number of tiny particles with mobility diameters (D_p) between 15 and 70 nm as compared with the “total” number of particles with mobility diameters between 15 and 800 nm. This percentage was calculated from full particle number size distributions ($10 < D_p < 800$ nm). Smallest particles ($10 < D_p < 15$ nm) were not considered because their measurement is affected by larger uncertainties (Wiedensohler et al., 2018), and by nucleation particle bursts. The range $15 < D_p < 70$ nm was selected as representative of particles emitted by primary sources (Giechaskiel et al., 2019; Giechaskiel, 2020; Ozgen et al., 2017; Tiwari et al., 2014). The second variable is the aerosol light absorption Ångström exponent (AAE). It represents the wavelength dependence of light absorption by aerosol particles. AAE values vary with particle sources and have commonly been used to apportion pollution particles between e.g. traffic and wood burning (Sandradewi et al, 2008). Traffic emitted particles (mainly from Diesel engines) have an AAE close to 1, while particles from wood combustion have more variable AAEs around 2 (Sandradewi et al, 2008). The mixture of pollution particles with primary or secondary aerosol from biogenic origin can also lead to AAE values much greater than 1. Since both variables are insensitive to air pollution dispersion, they are much less variable than the extensive variables (i.e. atmospheric concentrations) they are derived from (e.g. Putaud et al., 2014). The values expected for these so-called intensive variables were calculated as the arithmetic averages observed during the 2017-2019 period.

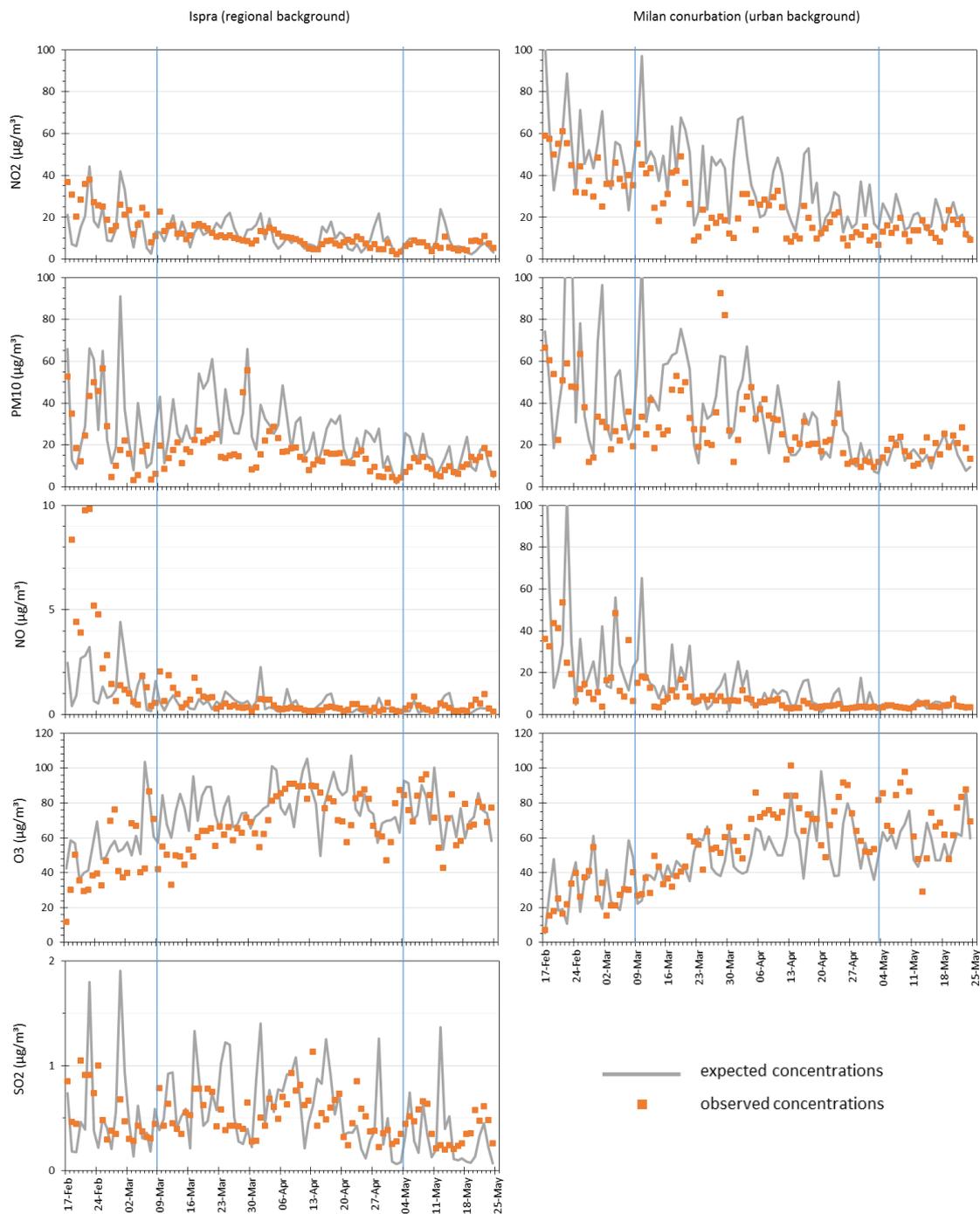


Figure 3: Observed (dots) and expected (lines) 2020 concentrations ($\mu\text{g m}^{-3}$) of NO_2 , PM_{10} , NO , O_3 , and SO_2 in Ispra (left hand side) and Milan conurbation (right hand side). Vertical lines indicate the beginning and end of the lockdown period.

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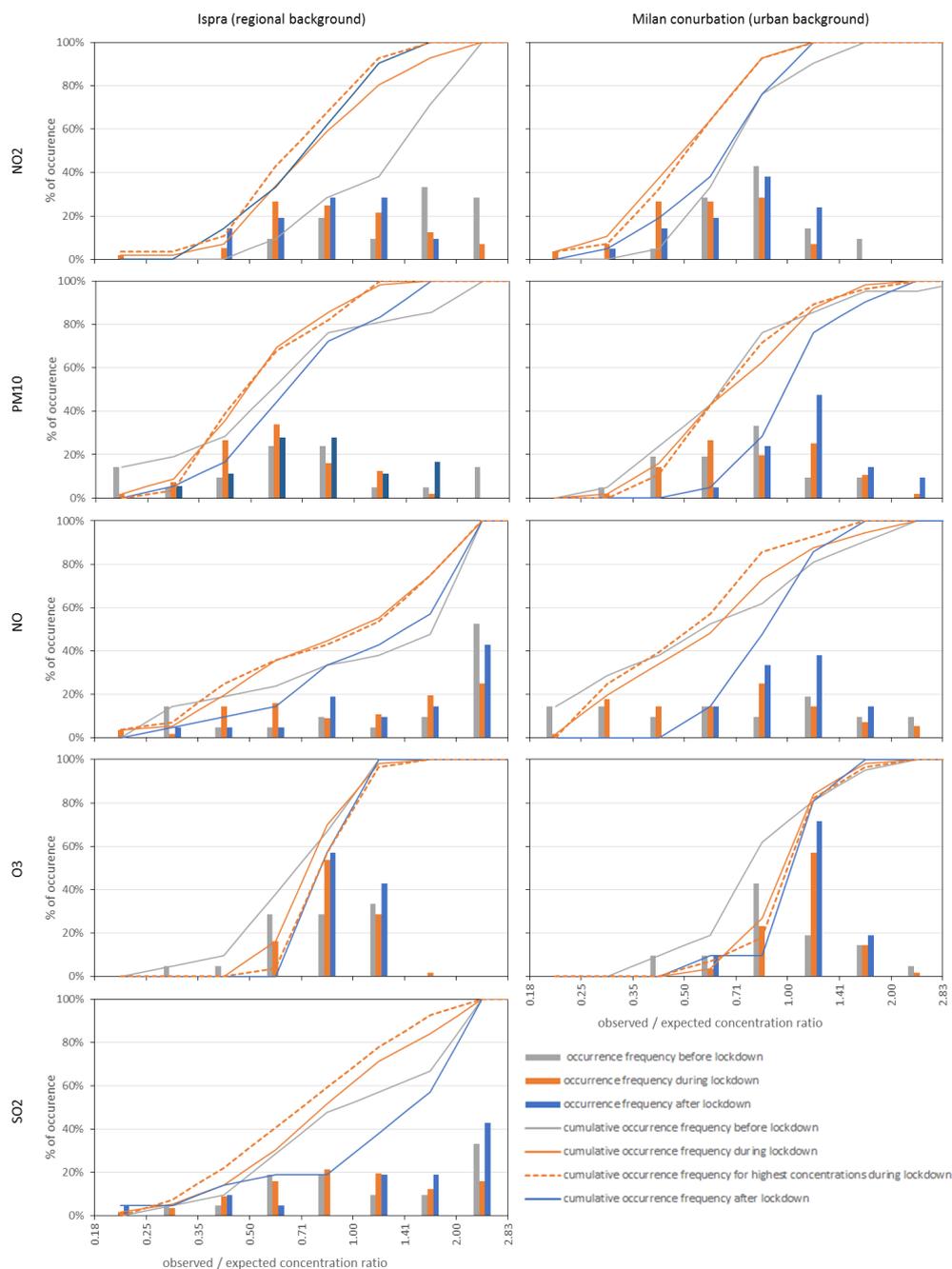


Figure 4: Occurrence frequency distributions of 2020 observed/expected concentration ratios (Obs / Exp) for NO₂, PM₁₀, NO, O₃ and SO₂ during the lockdown period, and during the 3 weeks before and after the lockdown period in Ispra (left) and Milan conurbation (right). Lines show cumulative frequencies of occurrence. Dashed lines show the cumulative frequency of occurrence of (Obs / Exp) ratios for the 28 days corresponding to the highest CAMS forecast values. NB: the last bin for NO in Ispra contains all values > 2.

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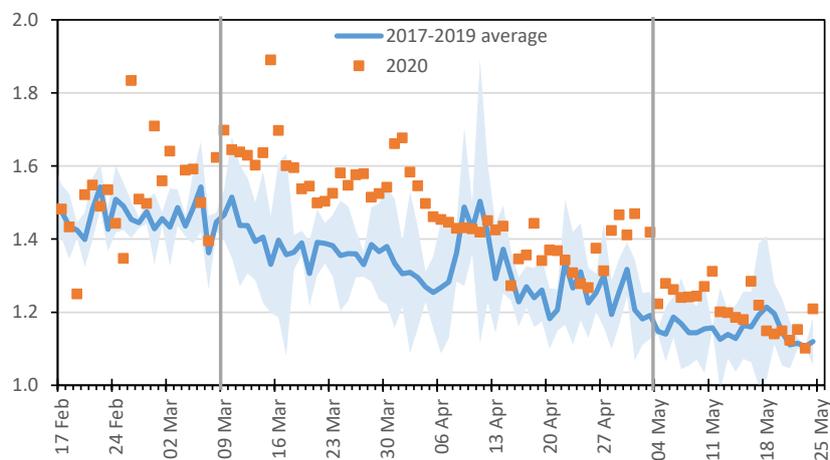


3 Results and discussion

The observation and CAMS-Ensemble forecast data used to estimate the values expected for the air pollution variables discussed in this section are described in Sections 1 and 2 of the Supplement to this article.

250 3.1. Regional background (Ispra)

The trend in AÅE observed in Ispra in 2017-2019 (and also in 2020) is consistent with a decreasing contribution of wood burning to particulate pollution from winter to summer. The AÅE values measured in 2020 can of course not be compared point to point to the 2017-2019 average in Figure 5 because the use of wood fuel for domestic heating also depends on weekend and cold evening occurrences. However, the clear increase in the AÅE average between 9 March and 4 May 2020 compared to the 3 weeks before, the 3 weeks after, and the corresponding period in 2017 - 2019 undoubtedly shows a change in particle sources related to lockdown measures (Table 1). A specific analysis focused on the 4 first weeks of the lockdown period (before significant amounts of biogenic aerosols are expected) suggest a - 45% reduction in aerosol from traffic (and a concomitant + 45% increase in aerosol from wood combustion) during that period.



260 *Figure 5: Aerosol light absorption Ångström exponent (AÅE) in 2020 (dots) compared to its 2017 – 2019 average (lines). The shaded area represents ± 1 standard deviation of the average. Vertical lines indicate the beginning and the end of the lockdown period.*

Particle number size distribution measurements in Ispra typically show modes at 25 – 50 nm during morning rush hours as well as in the evening in winter. Particle primary sources include fuel combustion by thermal engines and liquid (oil) or solid fuel (e.g. wood) combustion for domestic heating. The ultrafine mode diameters of primary particle emissions range from 50 to 100 nm for domestic heating (e.g. Tiwari et al, 2014; Ozgem et al., 2017), and range from 10 to 90 nm for engines (e.g. Giechaskiel et al, 2019; Giechaskiel, 2020). Measurements also show that peaks in the number of 15-70 nm particles can result from the growth of nucleation particles in the afternoon. The percentage of 15-70nm particles generally increased from mid-February till end of May in 2017 - 2019 (Figure 6). Considering that (1) wood burning combustion for domestic heating did not decrease during the lockdown period, (2) nucleation and growth of secondary aerosol particles were observed on sunny days during the lockdown



period from 6 April 2020, and (3) that mostly morning peaks in particle number diminished during the lockdown period especially from 11 March to 13 April 2020, the relative “disappearance” of 15 - 70 nm particles during the lockdown period (Figure 6) can be attributed to a decrease in traffic related to lockdown measures. Since the atmospheric lifetime of 15 - 70 nm particles is 3 - 12 hours, local to regional traffic was concerned. Although it significantly increased after 4 May 2020, the percentage of 15 - 70 nm particles did not get back to the level observed before the lockdown as lockdown measures were relaxed.

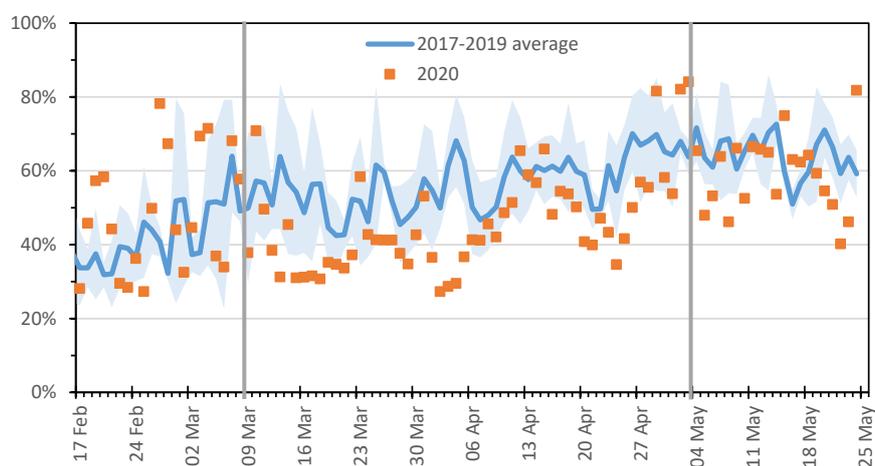


Figure 6: Percentage of sub-70 nm particles in 2020 (dots) compared to 2017 – 2019 average (line). The shaded area represents ± 1 standard deviation of the average. Vertical lines indicate the beginning and the end of the lockdown period.

The decrease in emissions from local traffic indicated by the drop in the percentage of the smallest particles (Figure 6) is the most probable cause for the decrease of NO related to the lockdown measures in Ispra (Figure 4). NO daily mean concentrations are indeed dominated by their morning peak corresponding to traffic rush hours (which disappears during weekends). During daytime, NO is rapidly converted to NO₂ and NO concentrations reach very low steady state values. Decreased NO emissions should therefore result in decreases in NO₂. Such a decrease in NO₂ (-40% on average) actually occurred in Ispra as a result of the lockdown measures from 9 March 2020 (Figure 4). In contrast, NO₂ did not “recover” from the lockdown measures, unlike NO of which concentrations increased again in comparison to expected concentrations as lockdown measures were relaxed on 4 May 2020 (Table 1). Due to its lifetime of about 1-2 days (Seinfeld and Pandis, 2016), NO₂ can travel rather long distances. Nitrogen oxides are also emitted by large combustion sources like thermal power plants, which also emit SO₂. However, our analysis of SO₂ data also reveals that sources of SO₂ that affect concentrations in Ispra decreased due to lockdown measures (Figures 3 and 4). The fact that NO₂ observed / expected ratios remained as low after, as during the lockdown period, could be explained by a slower increase in traffic on the regional scale as compared to the local scale.



Table 1: Observed / expected ratios for pollutant concentrations and aerosol characteristics before, during and after the lockdown measures in Ispra (regional background) and Milan (urban background).

	Ispra (Regional Background)				Milan (Urban Background)			
	before	during	during (> median)	after	before	during	during (> median)	after
NO ₂	1.52	0.89	0.78	0.82	0.83	0.57	0.59	0.73
PM ₁₀	0.67	0.60	0.61	0.77	0.77	0.83	0.81	1.20
NO	1.95	1.11	1.09	1.60	0.69	0.70	0.58	0.99
O ₃	0.79	0.87	0.96	0.98	0.93	1.17	1.22	1.15
SO ₂	1.19	1.02	0.85	1.59				
sub 70nm %	1.09	0.79	0.75	0.91				
AAE	1.04	1.12	1.14	1.05				

Regarding secondary pollutants, the highest O₃ concentrations significantly increased compared to expected concentrations during the lockdown period in comparison with the 3 weeks before (Figure 7 and 4). This suggests that O₃ peaks are usually diminished by NO titration during this period of the year in Ispra, and that the abatement in NO_x emissions revealed by NO and NO₂ data analyses led to a reduction of this effect. The relaxation of lockdown measures led to a further increase in O₃. Since O₃ production is generally VOC limited in May in Ispra, this increase in O₃ is probably due to an increase in anthropogenic emissions of VOCs from e.g. local traffic. In the case of PM₁₀, which is mainly composed of secondary particulate species in Ispra (Larsen et al., 2012), no significantly decrease compared to expected concentrations could be identified as lockdown measures were implemented (Figure 3 and 4). This is because the decrease in PM₁₀ related to traffic was compensated by the increase from wood burning for domestic heating, at least during the first half of the lockdown period. In contrast, PM₁₀ significantly but marginally increased as lockdown measures were relaxed on 4 May 2020 at a time of the year (from May onwards) where wood burning combustion for domestic heating is largely reduced.

3.2. Urban background (Milan conurbation)

To represent Milan urban background, we used data from 3 urban background sites located in Milan hinterland and in Milan city centre (Figure 1). NO₂ significantly decreased (-30% on average) compared to expected concentrations as lockdown measures were implemented (Figures 3 and 4). NO₂ significantly but not totally “recovered” when lockdown measures were relaxed (Table 1), which suggests that not all sources determining NO₂ concentrations in Milan were fully reopened. However, the increase in NO after the end of the lockdown period suggests that local traffic largely resumed. Perhaps NO_x emissions on a broader scale did not yet reach their usual intensity during the 3 first weeks after the end of the lockdown period, as already suggested by NO₂ data from Ispra. Regarding NO, it should be noticed that a significant decrease in comparison to the weeks before the lockdown period could only be detected at the city centre station (Figure S8). Both sites in the hinterland are much closer to highways and might reflect more NO emissions from heavy duty vehicles, whose traffic did not decrease that much at least during the first weeks of the lockdown period.

As in Ispra, the implementation of lockdown measures on 9 March 2020 led to increases of O₃ in Milan conurbation compared to expected concentrations (Figures 3 and 4). This can be explained by the decrease of O₃ titration by NO in a pollution regime where photochemical O₃ production is limited by the availability of volatile organic compounds. The relaxation of lockdown measures did not lead to the expected decrease in O₃ (Figure 4), perhaps because NO_x emissions did not fully recover during the 3 weeks following the end of the lockdown period.



Again, as in Ispra, no significant change in PM₁₀ could be detected when lockdown measures were implemented. This is very likely due to the fact that decreased emissions of PM₁₀ (and PM₁₀ precursors) were compensated by increases from other sources like domestic heating and enhanced formation of secondary PM. In particular, Huang et al. (2020) reported that increased oxidative capacities of the atmosphere (e.g. higher O₃ concentrations) resulted from the drastic reductions in NO_x emissions following from the lockdown measures in China, that in turn lead to an increase in the formation rate of nitric acid (HNO₃). Such a phenomenon in northern Italy, together with sustained emissions of ammonia from agriculture, which was not affected by the lockdown (ARPA Lombardia), could have resulted in increased formation of particulate ammonium nitrate (NH₄NO₃) and therefore an increase in PM₁₀ concentrations beyond expected concentrations in the Milan conurbation. For the 3 weeks following 4 May 2020, the relaxation of lockdown measures led to a further increase in PM₁₀ in comparison to expected concentrations in Milan. This might be attributed to the upturn in traffic and particularly to the re-suspension of dust from roads that had been little used for several weeks.

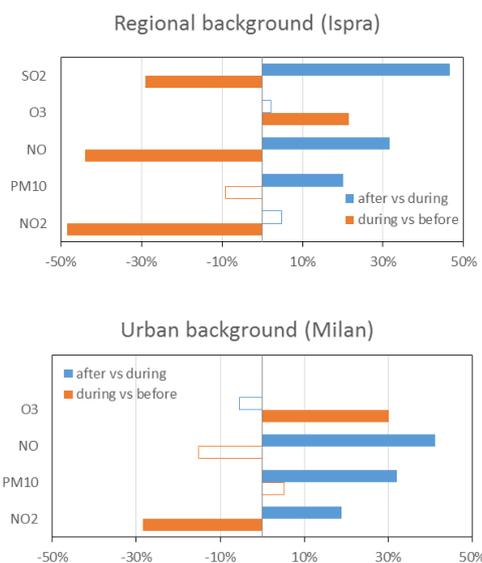


Figure 7: Changes in observed / expected concentration for the 3 weeks before and after the lockdown in comparison with the 28 days of lockdown corresponding to 50% highest CAMS daily forecasts for each pollutant. Filled bars represent statistically significant differences. Empty bars represent differences that are not significantly different from zero.

4 Conclusions

Northern Italy has been an air pollution hot spot for decades due to a high population density, intense economic activities, and particular orographic and meteorological conditions. Northern Italy also hosted the very first clusters of COVID-19 epidemic in Europe from February 2020, while containment measures were gradually implemented from the end of February with strict lockdown measures in force from 9 March to 4 May 2020. Lockdown measures impacted numerous economic sectors and activities with potential impacts on air pollution. Concentrations of several pollutants decreased from February to May 2020, as they have done every year for several decades, mainly due to seasonal variations in emissions and meteorological conditions. We



isolated the specific impact of lockdown measures on air pollution by comparing observed to expected data at one regional background site (Ispra) and 3 urban background sites (in Milan conurbation) for the period 17 February -24 May 2020. All 4
345 stations were in the COVID-19 “red zone”. Expected pollutant concentrations were derived from CAMS Ensemble forecasts, which are based on actual meteorological conditions and historical emissions estimates that ignored the COVID-19 epidemic and related lockdown measures. Changes in observed pollutant concentrations compared with computed expected concentrations for the lockdown period and the 3 weeks before and after the lockdown period therefore directly reflect the impact of lockdown measures on air pollution.

350 We showed that lockdown measures had statistically significant impacts on concentrations of most gaseous pollutants (Table 1). However, we were not able to highlight systematic effects on PM₁₀ concentrations, probably due to the wide variety of primary and secondary sources of PM₁₀.

Focusing on the effect of the lockdown measures on days for which the CAMS ensemble model forecast concentrations were above the median for the lockdown period (Figure 7), we found that NO₂ concentrations decreased by about -30% and -
355 50% at the urban and regional background sites, respectively, due to the lockdown implementation on 9 March 2020. This is consistent with similar decreases in NO concentrations. The relaxation of lockdown measures on 4 May led to a partial recovery in NO₂ concentrations in Milan (urban background), but not in Ispra (regional background). Unlike NO₂, PM₁₀ concentrations were not significantly affected by lockdown measures. The increase in the aerosol light absorption Ångström exponent observed in Ispra during the lockdown period in comparison with the same period in 2017 – 2019 indicates that this is because the decrease
360 in traffic-related PM₁₀ was compensated by an increase in PM₁₀ associated with wood burning for domestic heating. PM₁₀ concentrations in Milan are to a great extent influenced by PM₁₀ ‘non urban’ and ‘non traffic’ sources (Thunis et al., 2018), including the formation of secondary aerosol. Sustained regional background PM₁₀ concentrations and a modified HNO₃ production regime associated with continuing NH₃ emissions from agriculture could explain the lack of decrease in PM₁₀ resulting from the implementation of the lockdown measures in Milan too. In contrast, the relaxation of lockdown measures led to an
365 increase of PM₁₀ concentrations at both urban and regional background sites (+20% and + 15%, respectively) in May, when domestic heating is much reduced. Specific aerosol data from Ispra suggest that the impact of both local pollution sources (traced by the percentage of particles with diameters between 15 and 70 nm) and major industrial pollution sources (traced by SO₂ concentrations) on regional background air pollution decreased by -30 to -40% when lockdown measures were implemented, and at least partially got back to “normal” when lockdown measures were relaxed. Lastly, it can be pointed out that lockdown
370 measures led to an increase in the highest O₃ concentrations at both the urban and regional background sites when they were implemented.

The sad experience of the COVID-19 epidemic and subsequent lockdown measures shows that drastic changes in mobility and economic activity can lead to 0% (insignificant) to -30 % reductions in air pollution in urban background areas. These figures suggest that the abatement of air pollution down to levels that do not have adverse effects on human health in northern Italy
375 may require structural changes in energy production, domestic heating, agriculture and transport.

Data availability. Observation data from Ispra are available at <https://data.jrc.ec.europa.eu/collection/abcis> and <https://actris.nilu.no/>. Observation data from Milan are available at <https://www.arpalombardia.it/Pages/Aria/Richiesta-Dati.aspx>. Model forecast data for all sites are available at <https://ads.atmosphere.copernicus.eu/cdsapp#!/dataset/cams-europe-air-quality-forecasts?tab=form>
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