



# What can we learn about urban air quality with regard to the first outbreak of the COVID-19 pandemic?

## A case study from central Europe

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**Abstract.** Motor vehicle road traffic in central Budapest was reduced by approximately 50 % of its ordinary level for several weeks as a consequence of various limitation measures introduced to mitigate the first outbreak of the COVID-19 pandemic in 2020. The situation was utilised to assess the real potentials of urban traffic on air quality. Concentrations of NO, NO<sub>2</sub>, CO, O<sub>3</sub>, SO<sub>2</sub> and particulate matter (PM) mass, which are ordinarily monitored in cities for air quality considerations, aerosol particle number size distributions, which are not rarely measured continuously on longer runs for research purposes, and meteorological properties usually available were collected and jointly evaluated in different pandemic phases. The largest changes occurred over the severest limitations (partial lockdown in the Restriction phase from 28 March to 17 May 2020). Concentrations of NO, NO<sub>2</sub>, CO, total particle number ( $N_{6-1000}$ ) and particles with a diameter < 100 nm declined by 68 %, 46 %, 27 %, 24 % and 28 %, respectively, in 2020 with respect to the average reference year comprising 2017–2019. Their quantification was based on both relative difference and standardised anomaly. The change rates expressed as relative concentration difference due to relative reduction in traffic intensity for NO, NO<sub>2</sub>,  $N_{6-1000}$  and CO were 0.63, 0.57, 0.40 and 0.22 (%/%), respectively. Of the pollutants which reacted in a sensitive manner to the change in vehicle circulation, it is the NO<sub>2</sub> that shows the most frequent exceedance of the health limits. Intentional tranquillising of the vehicle flow has considerable potential for improving the air quality. At the same time, the concentration levels of PM<sub>10</sub> mass, which is the most critical

pollutant in many European cities including Budapest, did not seem to be largely affected by vehicles. Concentrations of O<sub>3</sub> concurrently showed an increasing tendency with lower traffic, which was explained by its complex reaction mechanism. Modelling calculations indicated that spatial gradients of NO and NO<sub>2</sub> within the city became further enhanced by reduced vehicle flow.

### 1 Introduction

The coronavirus disease (COVID-19) is caused by the novel, Severe Acute Respiratory Syndrome CoronaVirus 2 (SARS-CoV-2) virus. The outbreak was declared a pandemic by the WHO on 11 March 2020 (WHO, 2020). National governments and international agencies and organisations enacted widespread emergency actions for individuals, some professionals, communities and the public to reduce the risk of infection and to combat the plague. As a consequence of the implemented measures, road traffic in many cities worldwide was reduced in a substantial manner and for a considerable time interval. In parallel, lower concentrations of several air pollutants were reported from both satellite observations and in situ measurements (Conticini et al., 2020; Frontera et al., 2020; Keller et al., 2020; Lal et al., 2020; Le et al., 2020; Lee et al., 2020; Liu et al., 2020; Mahato et al., 2020; Morawska and Cao, 2020; Nakada and Urban, 2020; Petetin et al., 2020; Tobías et al., 2020; Wang et al., 2020).

This situation offers a unique possibility for atmospheric scientists to investigate experimentally some important atmospheric chemical and physical issues, including urban air quality and climate change under extraordinary conditions of lower traffic and industrial productivity (Sussmann and Rettinger, 2020). The results and consequences of this real “ambient experiment” can be utilised to determine the true potentials of action plans on tranquillising urban road circulation for handling air quality, overcrowding, traffic congestions, noise contamination and other environmental, health and climate impacts in large cities.

The task is, however, somewhat complicated. Actual concentrations of atmospheric constituents can depend on (1) their emissions from several sectors, (2) their physical removal processes, (3) local meteorological conditions, mainly precipitation ( $P$ ), wind speed (WS), planetary boundary layer height (PBLH) and atmospheric stability, (4) their (long-range) transport and (5) possible photochemical reactions, which are largely influenced by other meteorological properties such as global solar radiation (GRad), relative humidity (RH) and air temperature ( $T$ ) and by availability of and interactions with other chemical species present in the air. Many of the phenomena or properties listed are, in addition, interconnected and confounding, which further obscures the situation since they create an internally interacting environmental system.

Tropospheric residence time of constituents can also play a role under non-steady-state conditions (Harrison, 2018). As a result, atmospheric concentrations at a fixed site change both periodically and randomly (fluctuate) on daily, seasonal or annual scales. The variations are also linked to the geographical location and features of urban sites (de Jesus et al., 2019).

Source-specific markers generated by internal combustion engines or added on purpose into their fuel (e.g. Horvath et al., 1988; Gentner et al., 2017) or multivariate statistical methods (Hopke, 2016) can be applied to estimate the importance of vehicle traffic for air quality. These methods usually require advanced analytical methods to obtain data for specific species, which may not be available with a required time resolution or need a larger number of data, which can be constrained by duration of the time intervals of interest. Another possibility is to examine jointly the time series of multicomponent atmospheric data sets. This approach (described later in more detail) can be utilised retrospectively and is generally applicable in different cities in the world, which were affected by road traffic restrictions.

In Hungary, a state of emergency was introduced on 11 March 2020. It involved sequential closure of education institutes, the beginning of working from home and social distancing. It was followed by restrictions on movement. During this, residences could only be left with specified basic purposes, administrative centres, restaurants and touristic places were closed, distant travels were ceased, public parks were closed for long weekends and there were various time limita-

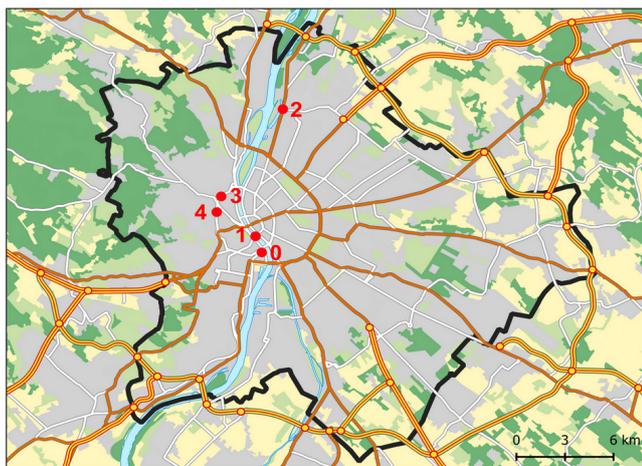
tions on shopping. The mitigating measures resulted in perceivable changes in vehicular road traffic and atmospheric concentrations. The main objectives of the present paper are (1) to introduce and demonstrate a general method for quantifying concentration changes, (2) to evaluate whether the changes observed were related to motor vehicle road traffic, (3) to assess the effect of traffic on these alterations, and (4) to estimate and debate the potentials of tranquillised urban vehicle flow for the air quality.

## 2 Methods

Criteria air pollutants, namely NO,  $\text{NO}_2 = \text{NO}_x - \text{NO}$ , CO,  $\text{O}_3$ ,  $\text{SO}_2$  and particulate matter (PM) mass in various size fractions, were involved in the study. The species originate from different sources. Vehicular road traffic is usually associated with NO and CO, while  $\text{NO}_2$  and  $\text{O}_3$  are formed by chemical reactions in the air. Contributions of residential heating, cooking, industrial activities, regional traffic in winter and secondary processes to  $\text{PM}_{2.5}$  mass are of large importance in many cities, including Budapest. At the same time,  $\text{PM}_{10}$  mass represents disintegration sources, e.g. wind-blown soil, crustal rock, mineral and roadside dust, resuspended dust by car movement, agricultural activities in the region, construction work and material wear such as tire abrasion of cars at kerbside sites (Salma and Maenhaut, 2006; Putaud et al., 2010; Harrison et al., 2012; Salma et al., 2020a). They all can be important, particularly under dry weather conditions.

Aerosol particle number concentrations in the diameter ranges from 6 to 1000 nm ( $N_{6-1000}$ ) and from 6 to 100 nm ( $N_{6-100}$ ) are mainly assigned to high-temperature emission sources (such as vehicle road traffic or incomplete burning) and atmospheric new particle formation and growth (NPF) events (Paasonen et al., 2016; Rönkkö et al., 2017; Salma et al., 2017). The latter process occurs as a daily phenomenon with a typical shape of its monthly occurrence frequency (Salma and Németh, 2019). This distribution changes in Budapest from year to year without any tendentious character (Salma et al., 2020b). Particles with a diameter from 25 to 100 nm ( $N_{25-100}$ ) in cities are mainly emitted by incomplete combustion or consist of grown new particles by condensation, while the size fraction with a diameter from 100 to 1000 nm ( $N_{100-1000}$ ) expresses physically and chemically aged particles; thus, they represent larger spatial extents (Salma et al., 2014; Mikkonen et al., 2020).

Approximate tropospheric residence times of  $\text{NO}_x$ , CO,  $\text{O}_3$ ,  $\text{SO}_2$  and PM are estimated to 1–2 d, 2 months, 1–2 months, 4–12 d and from several hours up to 1 week depending largely on particle size and chemical composition, respectively (Warneck and Williams, 2012; Harrison, 2018).



**Figure 1.** Location of the measurement sites in Budapest. 0: BpART Laboratory, 1: Szabadság Bridge, 2: Váci Road, 3: Széna Square and 4: Alkotás Road. The border of the city (in black colour), the Danube River and the major routes are also indicated.

## 2.1 Experimental data

The concentrations of NO / NO<sub>x</sub>, CO, O<sub>3</sub>, SO<sub>2</sub>, PM<sub>10</sub> mass and PM<sub>2.5</sub> mass were measured by chemiluminescence (Thermo 42C), IR absorption (Thermo 48i), UV fluorescence (Ysselbach 43C), UV absorption (Ysselbach 49C) and beta-ray attenuation (two Environment MP101M instruments with PM<sub>10</sub> and PM<sub>2.5</sub> inlets) methods, respectively, with a time resolution of 1 h. The concentrations of gases were expressed at a temperature of 293 K and a pressure of 101.3 kPa. The particle number concentrations were determined by a flow-switching type differential mobility particle sizer (DMPS; Salma et al., 2016b) with a time resolution of 8 min. The latter measurements were performed in a diameter range from 6 to 1000 nm in 30 size channels with equal width in the dry state of particles. The meteorological data of *T*, RH and WS and of GRad were measured by standardised sensors (HD52.3D17, Delta OHM, Italy, and SMP3 pyranometer, Kipp and Zonen, the Netherlands, respectively) with a time resolution of 1 min.

The DMPS and meteorological measurements were accomplished at the Budapest platform for Aerosol Research and Training (BpART) Laboratory (47°28′29.9″ N, 19°3′44.6″ E; 115 m above mean sea level) of the Eötvös University (Fig. 1). The location represents a well-mixed, average atmospheric environment for the city centre due to its geographical and meteorological conditions (Salma et al., 2016a). The local emissions include diffuse urban traffic exhaust, household/residential emissions and limited industrial sources together with some off-road transport (Salma et al., 2020a). In some time intervals, long-range transport of air masses can also play a role.

The data of the criteria air pollutants were acquired from a measurement station of the National Air Quality Network at Széna Square (Fig. 1) located 4.5 km from the BpART Laboratory in the upwind-prevailing direction (Salma and Németh, 2019). This station serves as a reference for our long-term air-quality-related research activities in several aspects and proved to be acceptable for this purpose.

Atmospheric transport of chemical species was assessed through large-scale weather types. We utilised macrocirculation patterns (MCPs), which were invented specifically for the Carpathian Basin (Péczeley, 1957; Maheras et al., 2018). The classification of the MCPs is based on the position, extension and development of cyclones and anticyclones relative to the Carpathian Basin considering the sea-level pressure maps constructed for 00:00 UTC in the North Atlantic–European region on a daily basis. A brief survey on the MCPs and the actual codes for year 2020 utilised in the interpretations are given in Table S1 and Fig. S1, respectively, in the Supplement. The relative occurrences of the weather types in year 2020 were roughly in line with multiple-year frequencies. Extended anticyclonic weather types usually indicate that the air masses are stagnant and that the importance of local or regional sources prevails over the air transport from distant sources. Under cyclonic weather conditions and frontal systems, the transported air masses can yield more pronounced effects and contributions.

A census of motor road vehicles was performed on three major routes and on a bridge over the Danube River by Budapest Public Roads Ltd. The measurement sites were on Szabadság Bridge, Váci Road, around Széna Square and Alkotás Road (Fig. 1), which are described in more detail in the Supplement. The counting was based on permanent electronic devices with inductive loops, and passenger cars, high- and heavy-duty vehicles and buses were recorded in both directions. The time resolution of the data was 1 h and their coverage was > 90 % of all possible items in a year. The sites cover a wide range of maximum hourly mean vehicle flow from about 1200 to 4600 h<sup>-1</sup>. Szabadság Bridge has the smallest traffic intensity of the sites, but it proved to be a very valuable microenvironment for the study since it is part of the internal boulevard. The routes showed coherent and common aggregate time properties and, therefore, their data are to be proportional to general vehicular traffic flow in the city centre.

## 2.2 Time intervals of interest

Time intervals from 1 January to 31 July in 2017, 2018, 2019 and 2020 were studied. This included all major measures related to the first outbreak of the COVID-19 pandemic in Budapest in 2020. Within these 7 months, five consecutive time intervals were selected for comparative purposes: (1) from 1 January till the beginning of the state of emergency at 15:00 on 11 March, which is referred to as the Pre-emergency phase, (2) from the beginning of the state of emergency to

27 March (till the beginning of the restriction on movement), which is called here the Pre-restriction phase, (3) from the beginning of the restriction on movement till its end in Budapest on 17 May, which is denoted the Restriction phase, (4) from the end of the restriction on movement in Budapest till the end of the state of emergency on 17 June, which is referred to as the Post-restriction phase, and (5) from the end of the state of emergency till 31 July, which is called the Post-emergency phase. An overview of the pandemic phases with further details of possible relevance for air quality issues is summarised in Fig. S1. Equivalent time intervals in year 2017–2019, which correspond to these phases, were considered for comparative purposes.

Local daylight saving time (LDST = UTC+1 or UTC+2) was chosen as the time base for the atmospheric concentrations and road traffic data because it was observed that the daily activity time patterns of inhabitants largely influence these variables in cities (Salma et al., 2014). The meteorological data were expressed in UTC+1 since their diurnal and seasonal behaviours are primarily controlled by Sun path and other natural processes.

### 2.3 Data treatment and modelling

Medical studies with the influenza virus indicated that absolute humidity (AH) constrains both transmission efficiency and virus survival more than RH (Shaman and Kohn, 2009). In order to facilitate the future comparison with other locations or cities in the world mainly for possible virology purposes, the hourly mean RH values (%) were converted to AH ( $\text{g m}^{-3}$ ) using a calculation recommended by WMO (2008):

$$\text{AH} = \frac{e(T_0) \times \exp\left(A \times \frac{T}{T+B}\right) \times \text{RH} \times C}{T + 273.15}, \quad (1)$$

where  $T$  is expressed in  $^{\circ}\text{C}$ ,  $e(T_0) = 6.112 \text{ hPa}$  is the saturation vapour pressure at  $T_0 = 0^{\circ}\text{C}$ ,  $A = 17.67$ ,  $B = 243.5^{\circ}\text{C}$  and  $C = 2.167$ . For air temperatures  $< 0^{\circ}\text{C}$ , we used an approximation for sub-cooled liquid water and adopted identical coefficients. This seems to be a plausible approach since the saturation vapour pressure curves for liquid water and ice surface follow each other closely near the freezing point. The AH values are summarised in Table S2, while we keep evaluating the RH because it seems to be more relevant for the purpose of this atmospheric study than the former property.

Vertical transfer of gases and aerosol particles emitted or generated at the Earth's surface can largely be affected by the dynamics of the PBLH. It is realised by the dilution of pollutants with mixing. The PBLH data were obtained from the 5th generation of the European Centre for Medium-Range Weather Forecasts (ECMWF) atmospheric reanalysis (ERA5) database using the Copernicus Climate Change Service (C3S, 2017). The ERA5 combines the modelled data of the ECMWF's Integrated Forecast System, version CY41R2 on 137 hybrid sigma vertical levels with newly available observations assimilated at every hour. In the present study, the

daily maximum PBLH values ( $\text{PBLH}_{\text{max}}$ ) were considered to be proportional to the volume of the mixed air parcel.

The data with a time resolution of smaller than 1 h were averaged for 1 h. The coverage of the hourly data was typically above 90 % of all items in each year. Descriptive statistics, thus count, minimum, median, maximum, and geometric mean with standard deviation (SD) of all variables, were derived for the time interval studied and each pandemic phase in year 2020 (Y2020). The characteristics were compared to the corresponding data in an average reference year (Y3Ref). This contains averages of the parallel hourly mean data of the year 2017–2019. Longer time spans than 3 years would not necessarily be advantageous since some chemical species in Budapest show tendentious change on a scale of 10 years (Mikkonen et al., 2020) and the urban traffic could also change substantially.

Comparative evaluations are often performed via the relative change (RDiff) of medians ( $m$ ) derived for a selected time span, which can be described as

$$\text{RDiff} = \frac{m(\text{Y2020}) - m(\text{Y3Ref})}{m(\text{Y3Ref})}. \quad (2)$$

In our case, the time spans considered were the intervals of the five pandemic phases in both Y2020 and Y3Ref. The quantity RDiff essentially expresses the ratio of medians. It is very important to stress immediately that the ratios are largely influenced by the absolute magnitude of variables and could be misleading if interpreted alone. In addition, different variables can have very different ranges of variability. A further metric that could, therefore, be involved is the standardised anomaly (SAly), which is described as

$$\text{SAly} = \frac{m(\text{Y2020}) - m(\text{Y3Ref})}{\text{SD}}. \quad (3)$$

This quantity expresses the observed differences in units of SD, so it brings out the relative asset of the actual difference. For GRad, which evolves daily from their very low values overnight in a large number, which were not considered, the anomaly was not standardised to its (expanded) SD, but instead it was calculated simply as a difference  $m(\text{Y2020}) - m(\text{Y3Ref})$  in its absolute unit.

A difference in the fluctuating and periodically varying data sets (see Sect. 1) over a pandemic phase in Y2020 was quantified to be significant with respect to the equivalent interval in Y3Ref if both their RDiff and SAly metrics were significant. The actual criteria adopted are specified and discussed in Sect. 3.5.

Average diurnal variations of all variables for workdays and holidays over each pandemic phase in the average reference year comprising 2017–2019 and year 2020 were calculated by selecting all individual data for a particular hour of day on workdays or on holidays over the time interval under evaluation and by averaging them.

The spatial distributions of the chemical species of interest over the city during each pandemic phase were modelled

via the surface concentrations derived from the Copernicus Atmosphere Monitoring Service (CAMS) with a grid resolution of  $0.1^\circ \times 0.1^\circ$  in order to study their potential differences (CAMS, 2019). The reanalysed concentrations are based on the following state-of-the-art European models: CHIMERE, EMEP, EURAD-IM, LOTOS-EUROS, MATCH, MOCAGE and SILAM (Marécal et al., 2015). The modelled concentrations are represented by the CAMS ensemble, which is the median of the available model results at each grid point. The CAMS modelling shares the meteorological driver of the ECMWF's Integrated Forecast System and the Monitoring Atmospheric Composition and Climate emission inventory of the Netherlands Organization for Applied Scientific Research. The system provides daily 96 h estimates with hourly outputs of several chemical species. The hourly analysis at the Earth's surface is done a posteriori for the past day using a selection of air quality data from the corresponding European monitoring stations.

### 3 Results and discussion

The changes in atmospheric concentrations are presented and interpreted after the effects of the confounding variability in local meteorological conditions and in (long-range) transport of atmospheric air masses are evaluated and quantified.

#### 3.1 Meteorological conditions

The hourly average meteorological data over the time interval considered were in line with ordinary characteristics measured at the BpART Laboratory (Salma and Németh, 2019; Mikkonen et al., 2020). The  $T$  in 2020 was colder by  $0.4^\circ\text{C}$  than in the average reference year, and the relative differences for median RH, WS, GRad and  $\text{PBLH}_{\text{max}}$  were  $-3\%$ ,  $-8\%$ ,  $+3\%$  and  $+15\%$ , respectively. These alterations, except for the  $\text{PBLH}_{\text{max}}$ , are not significant (they remained within  $\pm 10\%$ ). There were, however, two important alterations from the multiple-year weather situations. First, spring 2020 was extraordinarily dry; it was the third driest season since 1901. This can likely be related to multifactorial meteorological reasons. Between 14 March and 24 April, anticyclonic weather types prevailed in the Carpathian Basin almost continuously for 41 d (Fig. S1). After this interval, the weather type was mostly cyclonic but with northerly wind, which ordinarily brings dry and cold air masses to the Budapest area. These factors together resulted in the long and severe drought. Finally, it was followed by frequent, continued and spatially extended rains in June (Fig. S1). Secondly, the number of foggy hours (160) in January 2020 was more than 4 times larger than in the average reference year. This conclusion is based on measurements at the Budapest Liszt Ferenc International Airport.

An overview of the major meteorological data during the whole state of emergency interval (98 d) is summarised in

**Table 1.** Medians of the daily maximum planetary boundary layer height (km) in the average reference year comprising 2017–2019 (Y3Ref) and year 2020 (Y2020) together with their relative difference (RDiff) in % and their anomaly standardised to SD (SAly) over the five consecutive phases of the first COVID-19 outbreak.

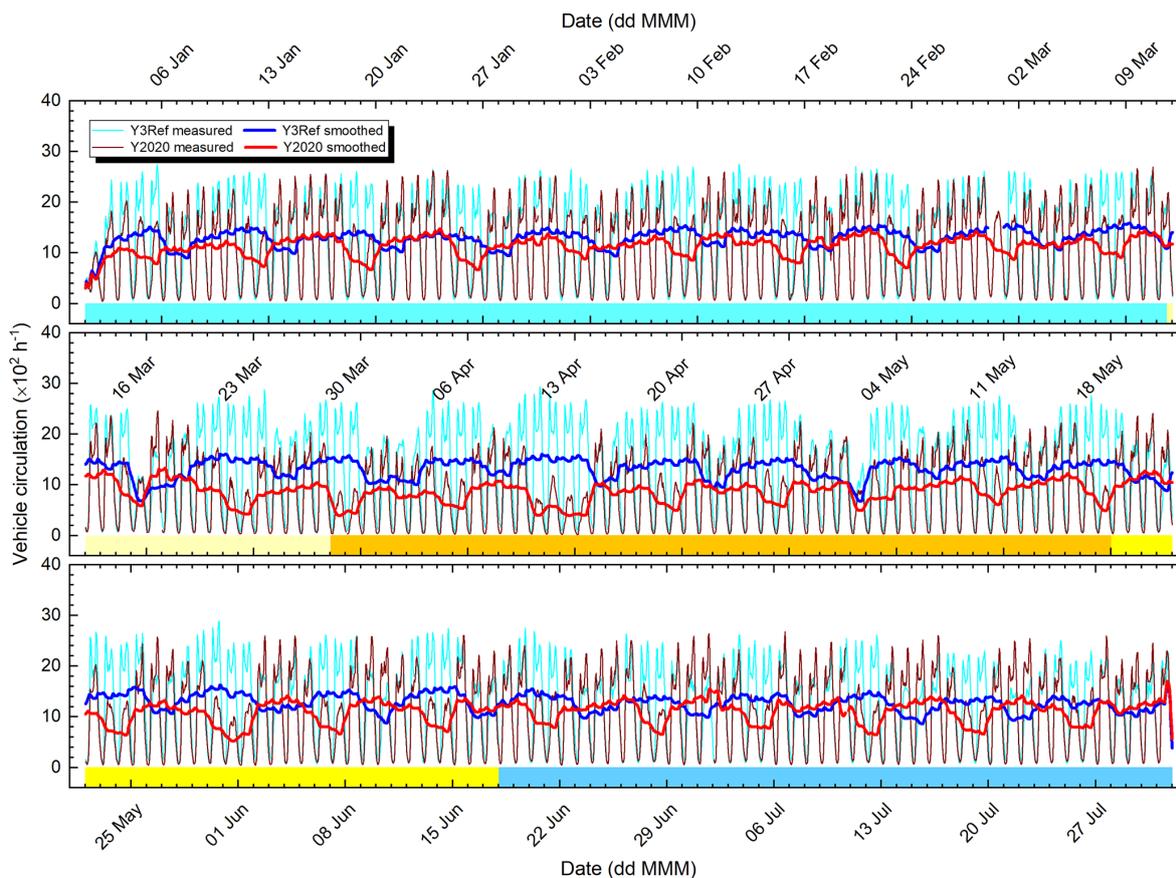
Pandemic phase	Y3Ref	Y2020	RDiff	SAly
Pre-emergency	0.66	0.88	+32	+0.4
Pre-restriction	1.4	1.4	+1	+0.0
Restriction	1.5	1.8	+18	+0.5
Post-restriction	1.6	1.3	-21	-0.7
Post-emergency	1.8	1.7	-8	-0.3

Table S2. The drought did not seem to influence substantially the WS and GRad but affected considerably the RH and indirectly the  $\text{PBLH}_{\text{max}}$ . The alterations in the  $\text{PBLH}_{\text{max}}$  in the average reference year and year 2020 over the pandemic phases are, therefore, quantified separately in Table 1 and are also displayed in Fig. S2. The time series for WS and  $T$  are also given in Figs. S3 and S4, respectively. It is seen in Fig. S2 that the Restriction phase – which is of particular interest for this study – was influenced by the  $\text{PBLH}_{\text{max}}$  in a more-or-less persistent manner without larger oscillations or fluctuations. The RDiff properties are taken into consideration when quantifying the concentration changes (Sect. 3.5). Lastly, it should also be mentioned that some of the differences in the meteorological data become small or insignificant when comparing them to their uncertainty intervals (in particular, for the modelled  $\text{PBLH}_{\text{max}}$ ).

#### 3.2 Motor vehicle road traffic

Time series of vehicle flow on a major route (Váci Road, site no. 2 in Fig. 1) over the time interval studied in the average reference year and year 2020 are shown in Fig. 2 as examples. The other urban sites exhibited very similar time behaviour and tendencies.

The time series for vehicle flow showed a clear periodicity. On each workday, two peaks – corresponding to the early morning and late afternoon rush hours – can be identified. In addition to this periodicity, the smoothed curves also revealed an obvious cycling due to a repeated workdays and holidays sequence. More importantly, the time series implied that in the Pre-emergency pandemic phase, the road traffic in the city centre in Y2020 was very similar to that in Y3Ref. The difference only appeared as a horizontal shift in time, which was caused by the occurrence of holidays in the average reference year and year 2020. Two weeks before the introduction of the restriction on movement, the vehicle circulation already started declining, and in the last week of the Pre-restriction phase, it already reached the level observed later in the Restriction phase. During these 8 or 9 weeks, the vehicular circulation on workdays was around the ordinary levels on holidays in 2017–2019. The circulation approached



**Figure 2.** Time series of motor vehicle circulation on a major route (Váci Road) in Budapest in both directions in the average reference year comprising 2017–2019 (Y3Ref) and year 2020 together with their 24 h smoothed curves over the five consecutive phases of the first COVID-19 outbreak. The phases are marked by the following colour codes: Pre-emergency phase lighter blue, Pre-restriction phase lighter yellow, Restriction phase orange, Post-restriction phase darker yellow and Post-emergency phase darker blue. The tick labels of the abscissa indicate the Mondays in 2020.

its ordinary values within or after the first week of the Post-restriction phase step-wisely. After that, the curves for the 2 years were at almost identical levels again. The changes in the vehicle flow are quantified in Sect. 3.5 together with the pollutant concentrations.

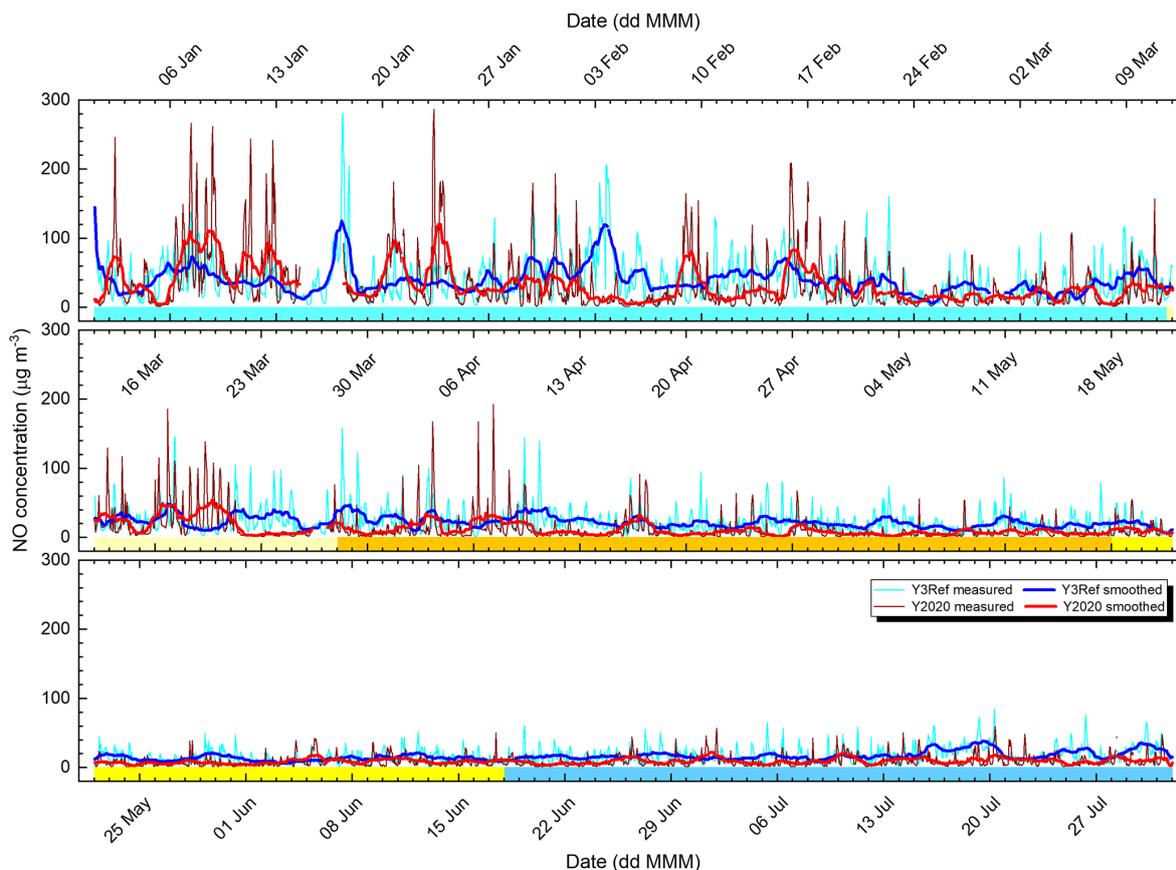
The shapes of the diurnal patterns (Fig. S5) for the average reference year and year 2020 were similar to each other, with some modifications. Differences could be identified in the Pre-restriction and Restriction pandemic phases between 16:00 and 19:00, when the traffic flow on weekends seemed to be systematically and in excess lower in 2020 than in the average reference year. This could be due to the limitations on shopping and to modified going-out routines of inhabitants under the restrictions. Similarly, the early morning peak on workdays in the Post-emergency (and partly Post-restriction) phases was smaller in excess in Y2020 than in Y3Ref, which can likely be linked to fewer people going physically to work due to propagated home-office jobs. To facilitate the comparison of diurnal patterns of vehicle circulation and of atmospheric concentrations, the plot show-

ing the diurnal variation of concentrations in the Restriction phase was extended by the vehicle flow in the same pandemic phase (Fig. 7).

### 3.3 Time series of concentrations

Time series of NO, O<sub>3</sub>, PM<sub>2.5</sub> mass and  $N_{6-1000}$  atmospheric concentrations over the time interval studied are shown in Figs. 3–6, respectively. The chemical species selected represent primary pollutant gases, secondary pollutant gases and two different aerosol properties, respectively. The corresponding curves for NO<sub>2</sub>, CO, SO<sub>2</sub>, PM<sub>10</sub> mass and  $N_{100-1000}$  are displayed in Figs. S6–S10, respectively.

The curves for both measured and smoothed data demonstrated that the concentrations varied substantially in time. The changes on the smoothed curves seemed to be fluctuations on a daily scale and, for some pollutants, they appeared to exhibit some tendencies on a monthly scale, while the data series possessed diurnal periodicity as well. The trends, i.e. the smoothed curves over the 7 months, are in line with the



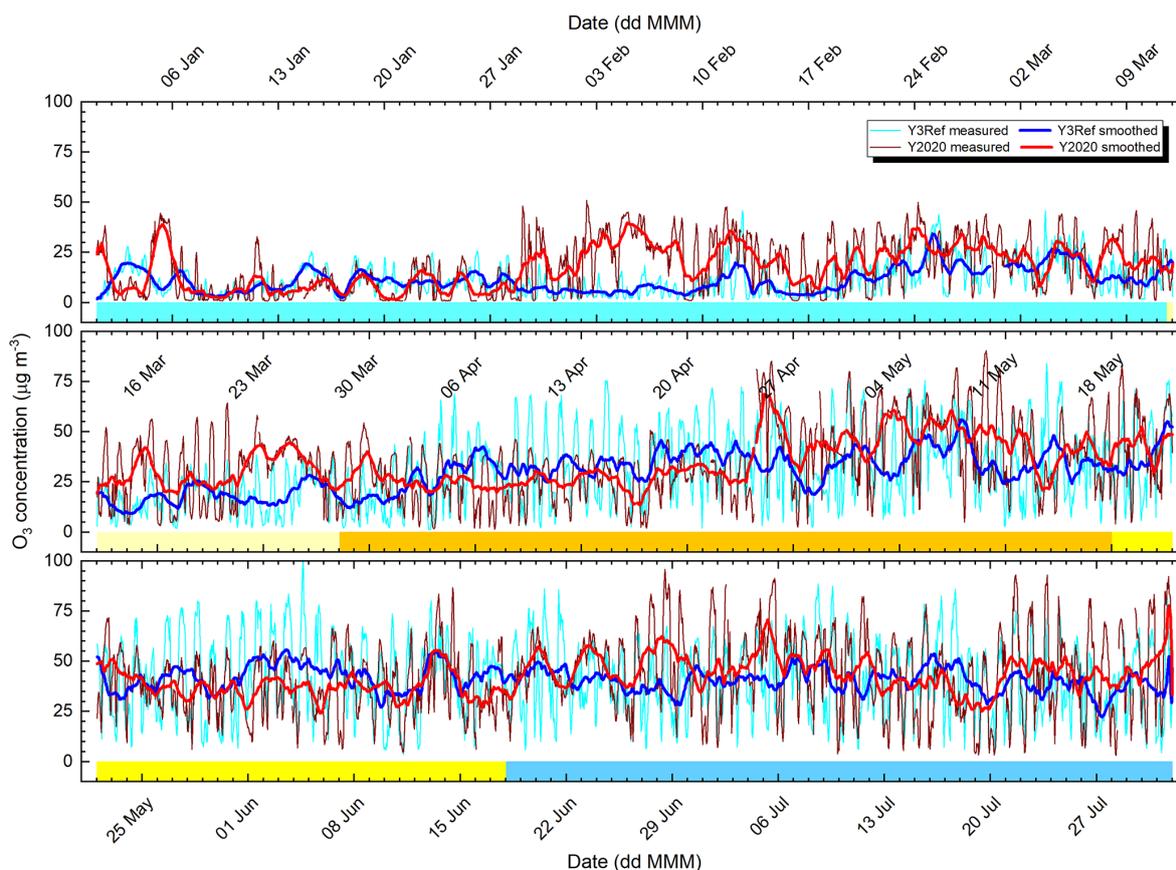
**Figure 3.** Time series of NO concentration in the average reference year comprising 2017–2019 (Y3Ref) and year 2020 together with their 24 h smoothed curves over the five consecutive phases of the first COVID-19 outbreak. The phases are marked by the following colour codes: Pre-emergency phase lighter blue, Pre-restriction phase lighter yellow, Restriction phase orange, Post-restriction phase darker yellow and Post-emergency phase darker blue. The tick labels of the abscissa indicate the Mondays in 2020.

distributions of the monthly median concentrations of the species at identical locations determined for several years (Salma et al., 2020b).

The annual relative SDs (RSDs) for NO, NO<sub>2</sub>, CO, O<sub>3</sub>, SO<sub>2</sub>, PM<sub>10</sub> mass, PM<sub>2.5</sub> mass,  $N_{6-1000}$ ,  $N_{6-100}$ ,  $N_{25-100}$  and  $N_{100-1000}$  in year 2017–2019 were 115 %, 56 %, 43 %, 91 %, 37 %, 56 %, 74 %, 63 %, 69 %, 68 % and 68 %, respectively (cf. Sect. 1). Their time distributions were complex. For species which do not normally show seasonal tendency such as particle number concentrations and perhaps PM<sub>10</sub> mass, the distributions of monthly RSDs were also featureless. For SO<sub>2</sub>, which tends to exhibit smaller concentration levels in summer than in winter, the distribution of its monthly RSDs seemed to have an opposite behaviour. For O<sub>3</sub>, which exhibits larger concentrations in summer than in winter, the distribution of monthly RSDs showed again an opposite behaviour. These relationships are in accordance with general metrological expectations. Excitingly, for NO, NO<sub>2</sub>, CO and perhaps PM<sub>2.5</sub> mass, the distributions of monthly RSDs appeared to roughly follow in parallel the concentration trends within the concentration ranges actually obtained.

The largest decrease in the RSDs from winter to summer was observed for NO, which was approximately 20 % (of its annual mean RSD). The latter association could likely be linked to meteorological conditions and source/sink intensities of these pollutants.

Many chemical species investigated originate from rather different sources. Nevertheless, their atmospheric concentrations often changed coherently, particularly in winter and early spring. A nice example is the interval of approximately 14–28 March 2020 when most species varied consistently. The MCPs for these days indicate strong anticyclonic weather types over the Carpathian Basin and stagnant and relatively calm meteorological conditions without precipitation in the area (Fig. S1). It is a confirmation of the common effects of regional meteorology on atmospheric concentrations and that the daily evolution of meteorology can have a higher influence on atmospheric concentrations than the source intensities under such specific conditions (Salma et al., 2020a). Its consequences for the air quality are discussed in Sect. 3.8.



**Figure 4.** Time series of  $\text{O}_3$  concentration in the average reference year comprising 2017–2019 (Y3Ref) and year 2020 together with their 24 h smoothed curves over the five consecutive phases of the first COVID-19 outbreak. The phases are marked by the following colour codes: Pre-emergency phase lighter blue, Pre-restriction phase lighter yellow, Restriction phase orange, Post-restriction phase darker yellow and Post-emergency phase darker blue. The tick labels of the abscissa indicate the Mondays in 2020.

The curves for  $\text{PM}_{2.5}$  mass and  $N_{6-1000}$  confirmed that there is a weak association between these two types of aerosol metrics (de Jesus et al., 2019). They are connected mainly via meteorological properties, which is anyway active for all pollutants. It was sensible, therefore, that both types of aerosol concentrations were included in the study as separate variables.

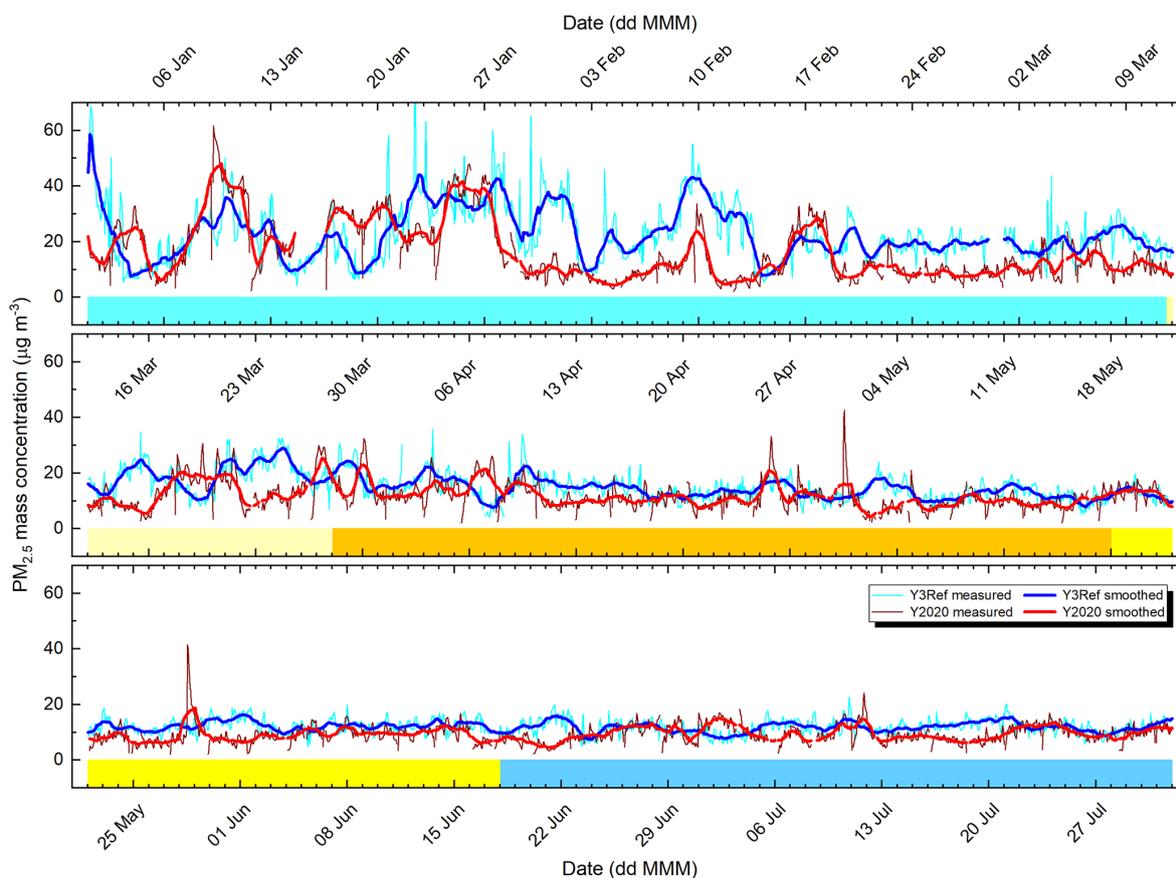
### 3.4 Diurnal variations

Average diurnal variations of  $\text{NO}$ ,  $\text{O}_3$ ,  $\text{SO}_2$ ,  $\text{PM}_{2.5}$  mass and  $N_{6-100}$  together with the vehicle circulation separately for workdays and holidays over the Restriction pandemic phase, for which the differences in the shapes are expected to be the largest, are shown in Fig. 7 as examples.

The curves of  $\text{NO}$  and  $N_{6-100}$  (together with  $\text{NO}_2$ ,  $\text{CO}$  and  $N_{6-1000}$ , which are not shown) followed the typical pattern of road traffic. They can largely be associated with vehicular sources (tailpipe emissions, primary and secondary particles) and can advantageously be applied for assigning potential concentration changes to traffic reduction. It is seen that their morning peak coincided with the peak of the morning rush

hour, while their evening peak appeared later than the peak of the afternoon rush hour. This shift could likely be related to the daily evolution and cycling of the PBLH and to mixing intensity. The curves of  $N_{6-100}$  contained in addition the characteristic midday peak, which is caused by atmospheric NPF events. It is worth realising that its position was shifted to a later time. There were only nine quantifiable NPF events during the Restriction phase in year 2020, which might not result in a representative shape. An alternative explanation could be that this peak was caused by the overlapping effects of direct traffic emissions and NPF events superimposed on each other. This experimental observation should definitely be investigated and clarified when the necessary data sets become available.

The curves for  $\text{O}_3$  seemed to be opposite to  $\text{NO}$  as far as both their daily variations and the orders of concentration magnitudes on workdays and holidays are concerned. These are in line with the understanding of their atmospheric processes and coupled reaction mechanisms (Jacob, 1999). In addition, the shapes in Y2020 seemed to be flattened with respect to Y3Ref. The  $\text{O}_3$  curves were also affected by the



**Figure 5.** Time series of  $\text{PM}_{2.5}$  mass concentration in the average reference year comprising 2017–2019 (Y3Ref) and year 2020 together with their 24 h smoothed curves over the five consecutive phases of the first COVID-19 outbreak. The phases are marked by the following colour codes: Pre-emergency phase lighter blue, Pre-restriction phase lighter yellow, Restriction phase orange, Post-restriction phase darker yellow and Post-emergency phase darker blue. The tick labels of the abscissa indicate the Mondays in 2020.

clock change, since the concentrations of  $\text{O}_3$  are substantially influenced by solar radiation.

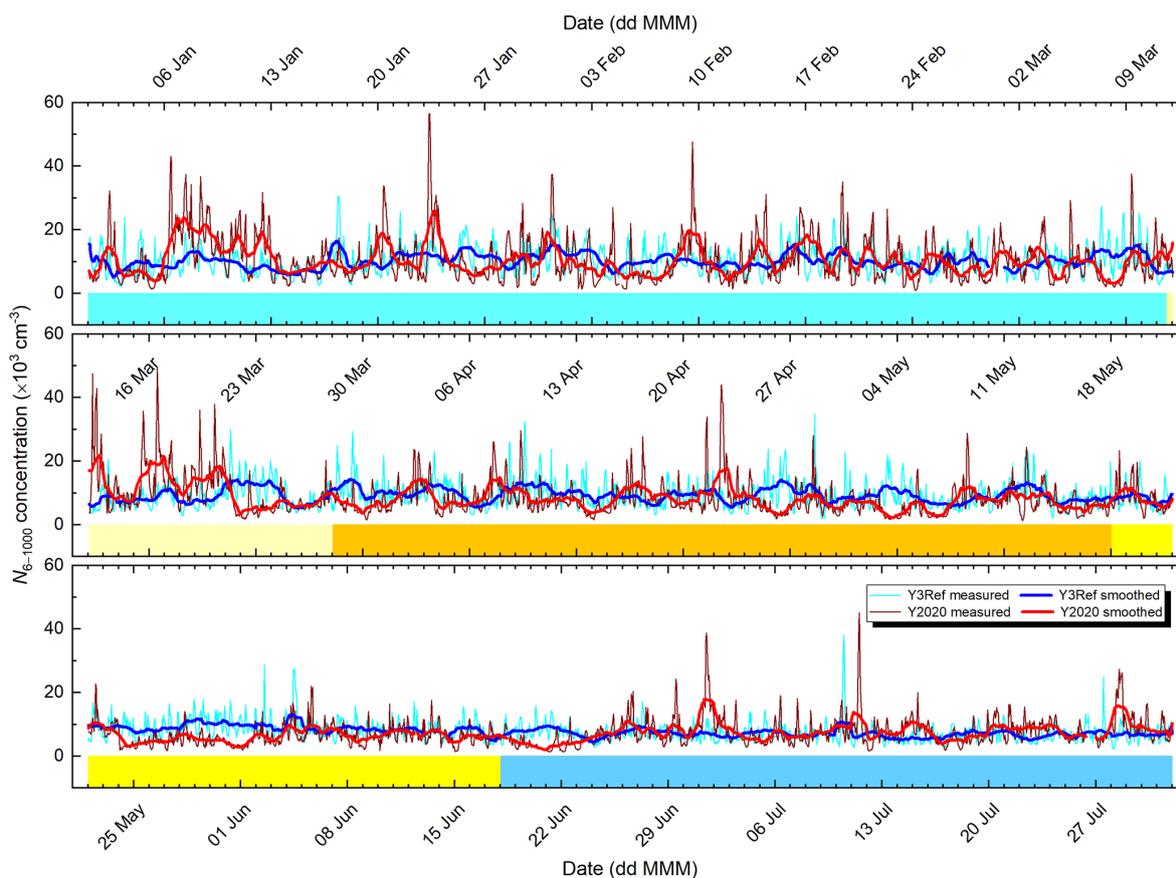
The curves for  $\text{SO}_2$  (together with  $\text{PM}_{10}$  mass and  $N_{100-1000}$ , which are not shown) tracked the traffic pattern very loosely if at all. They could partially be related to traffic through diesel fuel, resuspension of urban dust by moving vehicles, dispersion of road surfaces, (non-exhaust) emissions from material wear of moving parts of vehicles and growth/ageing of exhausted particles from vehicles (Salma and Maenhaut, 2006). Additional changes in the shape of the time variation of  $\text{SO}_2$  could be caused by altered heating of and cooking at homes due to the spreading practice of working from home.

There was no obvious connection between the traffic and  $\text{PM}_{2.5}$  mass, which confirms our earlier conclusion that the fine particles in Budapest mainly originate from non-vehicular sources (Salma et al., 2020a).

### 3.5 Quantification of concentration changes

There are several mathematical statistical tests to determine whether atmospheric concentrations over some time intervals in different years belong to the same distribution or not. These methods, however, quantify the joint influence of all environmental effects (Sect. 1) and do not provide information on their causal relationships. The method described and applied below allows us to unfold some potential confounding influence of environmental variables (e.g. PBLH) from concentration changes in order to gain a closer insight into the source intensities of motor vehicles.

Median concentrations of pollutant gases and aerosol particles, median traffic circulation data together with their relative differences and standardised anomaly values for the five pandemic phases in the average reference year and year 2020 are summarised in Tables 2–6. It should be noted that the standardised anomalies are rather small when recalling, for instance, the rigorous concept of the limits of detection ( $3 \times \text{SD}$ ) and determination ( $10 \times \text{SD}$ ) in analytical chemistry. This is largely caused by the strong dynamic features



**Figure 6.** Time series of  $N_{6-1000}$  concentration in the average reference year comprising 2017–2019 (Y3Ref) and year 2020 together with their 24 h smoothed curves over the five consecutive phases of the first COVID-19 outbreak. The phases are marked by the following colour codes: Pre-emergency phase lighter blue, Pre-restriction phase lighter yellow, Restriction phase orange, Post-restriction phase darker yellow and Post-emergency phase darker blue. The tick labels of the abscissa indicate the Mondays in 2020.

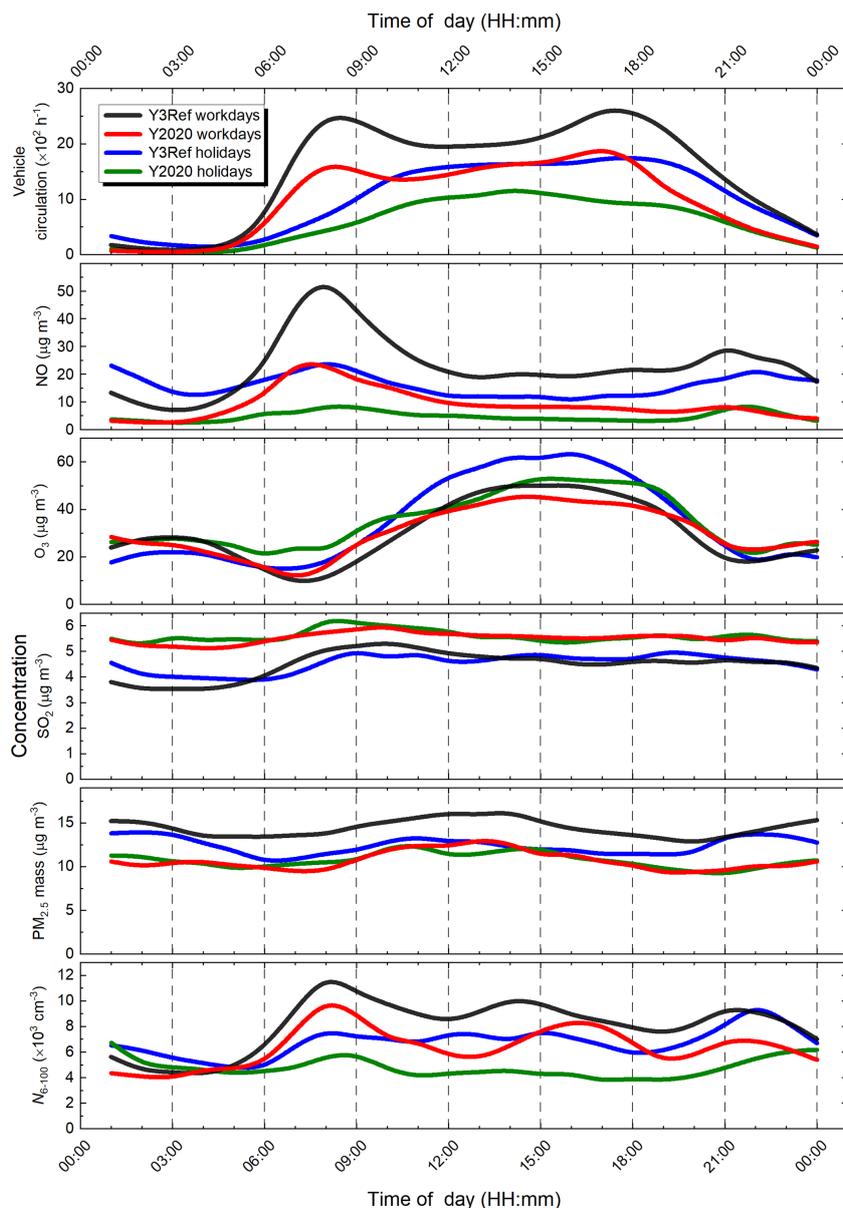
of related atmospheric properties and processes (Sects. 1 and 3.3).

We showed in Sect. 3.1 that it is the  $PBLH_{\max}$  of the meteorological conditions that likely caused the largest side effects on the concentrations, and, therefore, its influence was taken into account. A change in median concentrations for a pandemic phase was quantified to be significant if both its relative difference fell outside the band of  $[\pm 10 - f_{\text{mix}} \times \text{RDiff}(PBLH_{\max})] \%$  and its  $SA_{\text{ly}}$  was outside the range of  $\pm 0.3$ . The multiplication factor  $f_{\text{mix}}$  accounts for non-homogeneous mixing of pollutants within the boundary layer and for the effects of the daily PBLH evolution. It was roughly estimated to be approximately 0.5. Its negative sign expresses that atmospheric concentrations vary in a reciprocal manner with PBLH. The selected criteria were based upon exercises with the data in the individual years 2017, 2018 and 2019. The procedure represents a sensible and consequent approach, though alternative limits could also be set.

The Pre-emergency phase (Table 2) fitted completely into the heating season. The traffic flows in the city centre were identical, except for Váci Road, where they were somewhat

lower in Y2020 than in Y3Ref. This could be caused by some local traffic arrangements. The  $PBLH_{\max}$  increased by 32 % (Table 1), which is substantial and affected the concentrations. Most concentration changes were not significant. The exceptions were NO, O<sub>3</sub>, PM<sub>10</sub> mass and PM<sub>2.5</sub> mass, and the latter two exhibited the largest anomalies. These two species have multiple sources. Organic matter and elemental carbon, for instance, make up approximately 35 % of the PM<sub>2.5</sub> mass in winter (Salma et al., 2020a), and biomass burning is the major source of carbonaceous aerosol in this season, with an approximate relative contribution to the total carbon of 67 %. The share of fossil-fuel combustion is around 25 %. This all implies that PM<sub>2.5</sub> mass concentrations can fluctuate extensively and irregularly in the heating season due to the source intensities. The reductions could also be related to changes in further meteorological properties such as  $T$  (mild February 2020, Fig. S4) or larger WS that acted on a shorter timescale than the pandemic phase (Fig. S3).

The higher O<sub>3</sub> concentration could partly be associated with the lower concentrations of NO. Ozone exhibits a strong seasonal dependency (Salma et al., 2020b). Lower concen-



**Figure 7.** Average diurnal variations of motor vehicle road traffic in both directions on a major route (Váci Road) in Budapest and of NO, O<sub>3</sub>, SO<sub>2</sub>, PM<sub>2.5</sub> mass and N<sub>6–100</sub> concentrations separately for workdays and holidays in the average reference year comprising 2017–2019 (Y3Ref) and year 2020 during the Restriction phase of the first COVID-19 outbreak.

trations in winter and early spring can be easily disturbed by their non-linear chemistry and by high WS. The modest SAly for O<sub>3</sub> suggests that this considerable relative concentration increase was mostly a consequence of low levels of O<sub>3</sub> in winter. The case nicely demonstrates the strength of and requirement for the coupled utilisation of RDiff and SAly criteria. Furthermore, the main differences in the concentrations appeared sporadically in an isolated manner. In addition, there was no coherence among the traffic-related variables. Therefore, all significant variations were interpreted as

results of inter-annual variability in local meteorology, emissions and formation processes.

The Pre-restriction phase (Table 3) was rather short (16 d), and, therefore, its interpretation should be approached with special caution due to some issues in representativity. It was also completely part of the heating season, and the extreme drought in the Carpathian Basin in 2020 could also play a role. The PBLH<sub>max</sub> was almost identical in both years (Table 1). The concentrations of NO<sub>2</sub>, PM<sub>2.5</sub> mass and perhaps NO declined, while O<sub>3</sub> was enhanced. Excitingly, CO did not show a substantial decrease. The changes could be af-

ected by lower traffic during its last half/week (Fig. 2) and increased GRad.

The beginning one-third part of the Restriction phase (Table 4) fell in the heating season, and it was fully incorporated into the extremely dry weather season. The vehicle flows were reduced by approximately half uniformly at all urban locations. Concentrations of NO, NO<sub>2</sub>, CO, PM<sub>2.5</sub> mass, N<sub>6–1000</sub> and N<sub>6–100</sub> changed significantly, and they all declined. The alterations happened in a systematic or continuous manner in time (Figs. 2–6, S6 and S7). These species can be associated with vehicular road traffic, except for PM<sub>2.5</sub> mass, which is linked more to household and residential sources. At the same time, some other important pollutants such as N<sub>100–1000</sub> or SO<sub>2</sub> – which are typically related to larger spatial extent or region and which could, therefore, be influenced by meteorology – did not change significantly. Similar reductions were reported for other urban locations in the world (Keller et al., 2020; Lal et al., 2020; Le et al., 2020; Lee et al., 2020; Tobías et al., 2020). This all can be interpreted that the alterations in NO, NO<sub>2</sub>, CO, N<sub>6–1000</sub> and N<sub>6–100</sub> concentrations were primarily caused by the lower vehicular traffic intensity in the city and that the PBLH could also contribute by approximately 9% in an absolute sense (Table 1). The increased O<sub>3</sub> can be explained by its production primarily from volatile organic compounds (VOCs) and NO<sub>x</sub> in the VOC-limited chemical regime even under decreasing NO<sub>x</sub> conditions (Jacob, 1999; Lelieveld and Dentener, 2000). This regime is typical for many large cities, where the VOCs can involve, for instance, aromatics such as benzene and toluene, which largely originate from traffic sources.

In the Post-restriction phase (Table 5), the vehicle flow recovered step-wisely. The PBLH<sub>max</sub> in Y2020 decreased substantially relative to Y3Ref (Table 1). Most chemical species such as NO<sub>2</sub>, SO<sub>2</sub>, PM<sub>10</sub> mass, PM<sub>2.5</sub> mass, N<sub>6–1000</sub>, N<sub>6–100</sub>, N<sub>25–100</sub> and N<sub>100–1000</sub> exhibited significant changes. The list also included variables which characterise the region. At the same time, some typical vehicular-related species such as NO and CO – which are not really water-soluble – were not among them. Most significant changes showed decreasing tendency, except for SO<sub>2</sub>, which increased. The latter was caused by a continuously increasing SO<sub>2</sub> concentration (Fig. S8) recorded at the other air quality monitoring stations as well. The increase was likely caused as a perturbation by some local sources in the upwind direction from the city. This all suggests that the alterations were mainly produced by arrival of continued and spatially extended rains in the second half of the pandemic phase (Fig. S1). The precipitation washed out many chemical species from the urban and regional atmospheres. This time interval unambiguously demonstrated that the regional weather can cause similar modifications in atmospheric concentrations as substantially reduced (by 50%) urban traffic.

In the Post-emergency phase (Table 6), the traffic was at its ordinary level and there were no larger weather alterna-

**Table 2.** Median atmospheric concentrations of NO, NO<sub>2</sub> (both in units of  $\mu\text{g m}^{-3}$ ), CO ( $\text{mg m}^{-3}$ ), O<sub>3</sub>, SO<sub>2</sub>, PM<sub>10</sub> mass, PM<sub>2.5</sub> mass (all in  $\mu\text{g m}^{-3}$ ), N<sub>6–1000</sub>, N<sub>6–100</sub>, N<sub>25–100</sub>, N<sub>100–1000</sub> (all in  $10^3 \text{ cm}^{-3}$ ) and median vehicle road traffic ( $\text{h}^{-1}$ ) on Szabadság Bridge, Váci Road, Széna Square and Alkotás Road in the average reference year comprising 2017–2019 (Y3Ref) and year 2020 together with their relative difference (RDiff, %) and their anomaly standardised to SD (SALy) for the Pre-emergency phase of the first COVID-19 outbreak. Chemical species with significant change are shown in bold.

Variable	Y3Ref	Y2020	RDiff	SALy
<b>NO</b>	31	18	−43	−0.5
NO <sub>2</sub>	51	40	−22	−0.7
CO	0.74	0.58	−21	−0.8
<b>O<sub>3</sub></b>	9.4	16	+68	+0.3
SO <sub>2</sub>	5.5	5.4	−1	−0.0
<b>PM<sub>10</sub></b>	45	29	−36	−1.2
<b>PM<sub>2.5</sub></b>	21	12	−42	−1.2
N <sub>6–1000</sub>	9.5	8.8	−7	−0.2
N <sub>6–100</sub>	7.2	6.8	−6	−0.1
N <sub>25–100</sub>	3.5	3.1	−10	−0.2
N <sub>100–1000</sub>	2.2	1.7	−21	−0.5
Szabadság B.	676	640	−5	−0.1
Váci R.	1589	1299	−18	−0.4
Széna S.	1374	1437	+5	+0.1
Alkotás R.	2517	2425	−4	−0.1

**Table 3.** Median atmospheric concentrations of NO, NO<sub>2</sub> (both in units of  $\mu\text{g m}^{-3}$ ), CO ( $\text{mg m}^{-3}$ ), O<sub>3</sub>, SO<sub>2</sub>, PM<sub>10</sub> mass, PM<sub>2.5</sub> mass (all in  $\mu\text{g m}^{-3}$ ), N<sub>6–1000</sub>, N<sub>6–100</sub>, N<sub>25–100</sub>, N<sub>100–1000</sub> (all in  $10^3 \text{ cm}^{-3}$ ) and median vehicle road traffic ( $\text{h}^{-1}$ ) on Szabadság Bridge, Váci Road, Széna Square and Alkotás Road in the average reference year comprising 2017–2019 (Y3Ref) and year 2020 together with their relative difference (RDiff) in % and their anomaly standardised to SD (SALy) for the Pre-restriction phase of the first COVID-19 outbreak. Chemical species with significant change are shown in bold.

Variable	Y3Ref	Y2020	RDiff	SALy
NO	20	12	−39	−0.3
<b>NO<sub>2</sub></b>	46	38	−18	−0.5
CO	0.60	0.56	−8	−0.2
<b>O<sub>3</sub></b>	18	33	+80	+0.7
SO <sub>2</sub>	5.1	5.4	+7	+0.3
PM <sub>10</sub>	34	30	−12	−0.3
<b>PM<sub>2.5</sub></b>	19	13	−32	−0.8
N <sub>6–1000</sub>	8.1	8.4	+4	+0.1
N <sub>6–100</sub>	6.7	6.9	+4	+0.1
N <sub>25–100</sub>	2.9	3.2	+9	+0.1
N <sub>100–1000</sub>	1.4	1.6	+11	+0.2
Szabadság B.	652	417	−36	−0.8
Váci R.	1522	939	−38	−0.7
Széna S.	1371	1001	−27	−0.5
Alkotás R.	2792	1925	−31	−0.7

**Table 4.** Median atmospheric concentrations of NO, NO<sub>2</sub> (both in units of  $\mu\text{g m}^{-3}$ ), CO ( $\text{mg m}^{-3}$ ), O<sub>3</sub>, SO<sub>2</sub>, PM<sub>10</sub> mass, PM<sub>2.5</sub> mass (all in  $\mu\text{g m}^{-3}$ ),  $N_{6-1000}$ ,  $N_{6-100}$ ,  $N_{25-100}$ ,  $N_{100-1000}$  (all in  $10^3 \text{ cm}^{-3}$ ) and median vehicle road traffic ( $\text{h}^{-1}$ ) on Szabadság Bridge, Váci Road, Széna Square and Alkotás Road in the average reference year comprising 2017–2019 (Y3Ref) and year 2020 together with their relative difference (RDiff) in % and their anomaly standardised to SD (SAlly) for the Restriction phase of the first COVID-19 outbreak. Chemical species with significant change are shown in bold.

Variable	Y3Ref	Y2020	RDiff	SAlly
<b>NO</b>	19	6.0	−68	−0.5
<b>NO<sub>2</sub></b>	44	26	−39	−1.1
<b>CO</b>	0.58	0.43	−27	−0.8
O <sub>3</sub>	31	35	+13	+0.2
SO <sub>2</sub>	5.4	5.5	+3	+0.1
PM <sub>10</sub>	32	28	−13	−0.3
<b>PM<sub>2.5</sub></b>	14	11	−22	−0.4
<b><math>N_{6-1000}</math></b>	8.8	6.7	−24	−0.5
<b><math>N_{6-100}</math></b>	7.4	5.3	−28	−0.6
$N_{25-100}$	3.2	2.8	−12	−0.2
$N_{100-1000}$	1.3	1.2	−5	+0.1
Szabadság B.	689	318	−54	−1.2
Váci R.	1626	803	−51	−1.0
Széna S.	1537	844	−45	−1.0
Alkotás R.	3031	1516	−50	−1.1

**Table 5.** Median atmospheric concentrations of NO, NO<sub>2</sub> (both in units of  $\mu\text{g m}^{-3}$ ), CO ( $\text{mg m}^{-3}$ ), O<sub>3</sub>, SO<sub>2</sub>, PM<sub>10</sub> mass, PM<sub>2.5</sub> mass (all in  $\mu\text{g m}^{-3}$ ),  $N_{6-1000}$ ,  $N_{6-100}$ ,  $N_{25-100}$ ,  $N_{100-1000}$  (all in  $10^3 \text{ cm}^{-3}$ ) and median vehicle road traffic ( $\text{h}^{-1}$ ) on Szabadság Bridge, Váci Road, Széna Square and Alkotás Road in the average reference year comprising 2017–2019 (Y3Ref) and year 2020 together with their relative difference (RDiff) in % and their anomaly standardised to SD (SAlly) for the Post-restriction phase of the first COVID-19 outbreak. Chemical species with significant change are shown in bold.

Variable	Y3Ref	Y2020	RDiff	SAlly
NO	12	6.4	−44	−0.2
<b>NO<sub>2</sub></b>	40	26	−35	−0.9
CO	0.48	0.42	−13	−0.3
O <sub>3</sub>	42	37	+11	−0.2
<b>SO<sub>2</sub></b>	4.7	5.9	+26	+1.0
<b>PM<sub>10</sub></b>	29	21	−28	−0.6
<b>PM<sub>2.5</sub></b>	12	9.3	−24	−0.4
<b><math>N_{6-1000}</math></b>	8.2	6.0	−27	−0.5
<b><math>N_{6-100}</math></b>	6.8	4.9	−27	−0.5
<b><math>N_{25-100}</math></b>	3.3	2.4	−27	−0.5
<b><math>N_{100-1000}</math></b>	1.3	1.0	−22	−0.3
Szabadság B.	670	575	−14	−0.3
Váci R.	1536	1137	−26	−0.5
Széna S.	1540	1387	−10	−0.2
Alkotás R.	2597	2281	−12	−0.2

**Table 6.** Median atmospheric concentrations of NO, NO<sub>2</sub> (both in units of  $\mu\text{g m}^{-3}$ ), CO ( $\text{mg m}^{-3}$ ), O<sub>3</sub>, SO<sub>2</sub>, PM<sub>10</sub> mass, PM<sub>2.5</sub> mass (all in  $\mu\text{g m}^{-3}$ ),  $N_{6-1000}$ ,  $N_{6-100}$ ,  $N_{25-100}$ ,  $N_{100-1000}$  (all in  $10^3 \text{ cm}^{-3}$ ) and median vehicle road traffic ( $\text{h}^{-1}$ ) on Szabadság Bridge, Váci Road, Széna Square and Alkotás Road in the average reference year comprising 2017–2019 (Y3Ref) and year 2020 together with their relative difference (RDiff) in % and their anomaly standardised to SD (SAlly) for the Post-emergency phase of the first COVID-19 outbreak. Chemical species with significant change are shown in bold.

Variable	Y3Ref	Y2020	RDiff	SAlly
<b>NO</b>	16	7.4	−54	−0.3
<b>NO<sub>2</sub></b>	39	27	−31	−0.7
CO	0.43	0.42	−2	−0.1
<b>O<sub>3</sub></b>	39	46	+17	+0.3
SO <sub>2</sub>	4.0	4.3	+9	+0.3
PM <sub>10</sub>	26	22	−15	−0.3
<b>PM<sub>2.5</sub></b>	12	9.1	−22	−0.3
$N_{6-1000}$	6.7	6.7	+0	+0.0
$N_{6-100}$	5.5	5.4	−2	−0.0
$N_{25-100}$	2.7	2.8	+5	+0.1
$N_{100-1000}$	1.1	1.2	+9	+0.1
Szabadság B.	690	663	−4	−0.1
Váci R.	1471	1218	−17	−0.3
Széna S.	1594	1511	−5	−0.1
Alkotás R.	2507	2531	+1	+0.0

tions. Most concentrations – including some major vehicle-related pollutants such as CO and  $N_{6-100}$  – did not change significantly. The exceptions were NO, NO<sub>2</sub>, O<sub>3</sub> and PM<sub>2.5</sub> mass. The first three variables are connected to each other through atmospheric chemistry. The changes can likely be linked to inter-annual variability in sources, sinks, meteorological properties that act on a shorter timescale than the pandemic phase and atmospheric transformation and transport – similarly to that observed in the Pre-emergency phase.

### 3.6 Change rates

Linear regression analysis between the median RDiff for vehicle traffic on one side and RDiff for pollutants corrected for the RDiff(PBLH<sub>max</sub>) on the other side for all pandemic phases yielded change rates and SDs for NO, NO<sub>2</sub>,  $N_{6-1000}$  and CO of  $0.63 \pm 0.23$ ,  $0.57 \pm 0.14$ ,  $0.40 \pm 0.17$  and  $0.22 \pm 0.08$ , respectively. For PM<sub>10</sub> mass and PM<sub>2.5</sub> mass, the rates were slightly negative and insignificant. The data points for the Post-restriction phase – which were substantially affected by precipitation and frontal weather systems – were excluded from this analysis.

The change rates suggest that nitrogen oxides vary sensitively with traffic, total particle number concentration shows considerable dependency, while variation of CO is modest. This is linked to their residence times as well. The PM mass

concentrations do not appear to be closely related to traffic intensity in central Budapest.

### 3.7 Spatial gradients

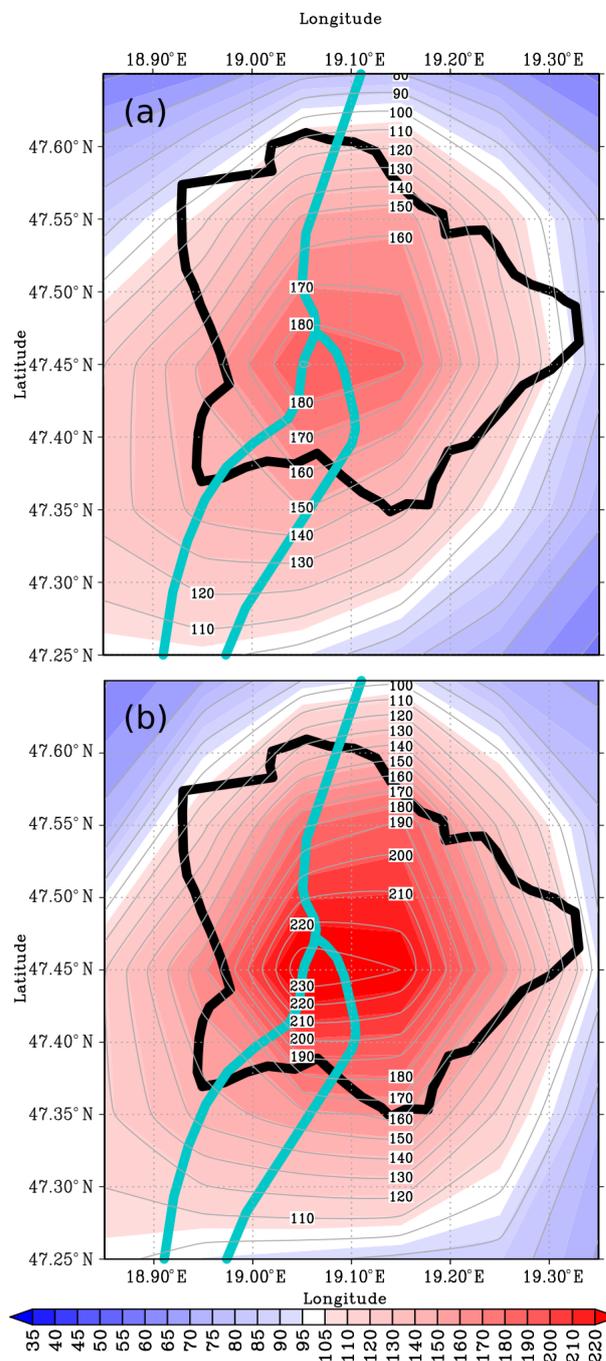
Spatial distributions of NO and O<sub>3</sub> derived by CAMS ensemble reanalysis in 2018–2019 and 2020 during the Restriction pandemic phase are shown in Figs. 8 and 9 as examples. The absolute concentrations can be different from the measured values due to the specialities in the applied models, while the relative tendencies are expected to be expressed correctly. Figure 8 indicates that the differences from the corresponding median (spatial gradients) in 2020 were larger than in 2018–2019. This can be explained if the relative concentration changes at the outer parts of the city or near-city background were even larger than in the centre. The spatial distribution of NO<sub>2</sub> was similar to NO, although its gradients were smaller than for NO. Spatial distributions of CO and PM<sub>2.5</sub> mass were featureless and similar to each other in 2018–2019 and 2020.

Spatial distributions of O<sub>3</sub> (Fig. 9) and, perhaps, SO<sub>2</sub> (which is not shown) exhibited relative decrease in the centre, whose gradients were relatively small and similar to each other for both time intervals. This all is in line with the tendencies observed in their measured concentrations (Sect. 3.3 and 3.4). We are aware that several pollutants originate from diffusive line sources, which can be enriched along roads and, therefore, much larger concentration gradients can occur on smaller spatial scales.

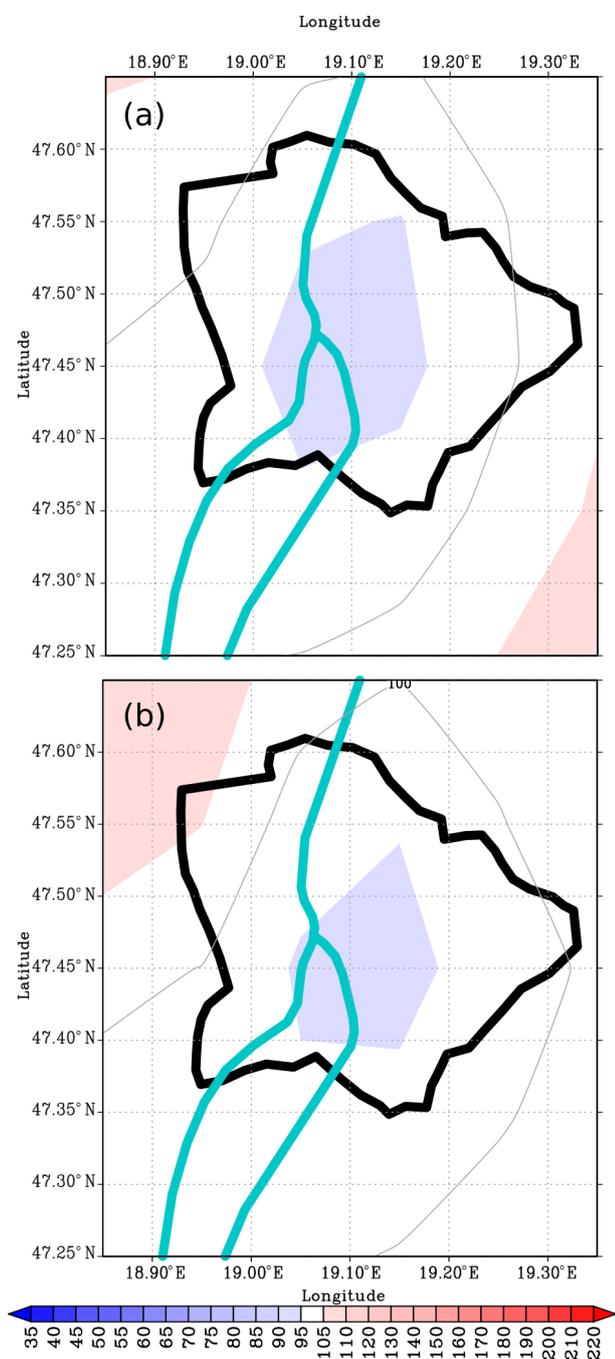
### 3.8 Potentials for improving air quality

In order to assess the importance of concentration changes that could be achieved by tranquilising the vehicle road traffic in Budapest, the atmospheric concentrations and their possible decrements were compared to various limit values (EU Directives, 2008; VM 4, 2011).

NO<sub>2</sub> exhibited the most frequent exceedances of standards for the protection of health. Its concentrations in 2017, 2018 and 2019 were larger in 172, 155 and 61 cases than the 1 h national health limit of 100 µg m<sup>-3</sup> (Fig. S6). The permitted number of exceedances is 18 a year. It is mentioned that this concentration limit is 200 µg m<sup>-3</sup> for the EU, which would be fulfilled completely. The NO<sub>2</sub> excess usually remained modest – particularly when contrasted with the smog alert thresholds of 350 (for the warning state) and 400 µg m<sup>-3</sup> (for the alarm state). The daily health limit of 85 µg m<sup>-3</sup> was also exceeded in 6, 2 and 0 d in the 3 years, respectively. A reduction of 10 % in vehicle circulation (that corresponds to a 6 % decline in NO<sub>2</sub> concentration) would decrease the number of exceedances of the 1 h national health limit to 114, 98 and 42, respectively (thus, typically by ca. 30 %), while the number of days above the daily health limit would be lowered to 2, 0 and 0, respectively.



**Figure 8.** Spatial distribution of median NO concentration in Budapest in 2018–2019 (a) and 2020 (b) during the Restriction phase of the first COVID-19 outbreak obtained from CAMS ensemble reanalysis. The concentrations were normalised to the overall spatial median concentrations of 0.93 and 0.59 µg m<sup>-3</sup>, respectively. The border of the city and the Danube River are indicated with curves in black and blue colours, respectively, for better orientation.



**Figure 9.** Spatial distribution of median  $\text{O}_3$  concentration in Budapest in 2018–2019 (a) and 2020 (b) during the Restriction phase of the first COVID-19 outbreak obtained from CAMS ensemble reanalysis. The concentrations were normalised to the overall spatial median concentrations of  $60$  and  $67 \mu\text{g m}^{-3}$ , respectively. The border of the city and the Danube River are indicated with curves in black and blue colours, respectively, for better orientation.

Less frequent though more severe exceedances than for  $\text{NO}_2$  happened for  $\text{PM}_{10}$  mass (Fig. S9). The daily mean  $\text{PM}_{10}$  mass concentrations in 2017, 2018 and 2019 exceeded the daily health limit of  $50 \mu\text{g m}^{-3}$  in 36, 93 and 57 d, respectively, at this actual air quality monitoring station. Most exceedances occurred in the heating season. Their permitted number is 35 a year. The smog alert thresholds for the warning and alarm states are  $75$  and  $100 \mu\text{g m}^{-3}$ , respectively. The numbers of exceedances for the warning stage were 12, 17 and 9, respectively. It should be added that smog alerts are announced on the basis of a complex set of conditions which include larger numbers of monitoring stations and days. As a matter of fact, the warning state was announced three times for 11 d in total and once for 2 d in 2017 and 2018, respectively. There was no smog alert in 2019. It is stressed that all alarm states since 2007 were announced exclusively because of high  $\text{PM}_{10}$  mass concentrations, and all alert intervals were confined to winter. This points to the role of local and regional meteorology and other sources than vehicle traffic (Salma et al., 2020a). There is a cold air pool that develops from time to time above the Carpathian Basin in winter, which generates a lasting  $T$  inversion and a shallow planetary boundary layer, restricts the vertical mixing and results in poor air quality over extended areas of the basin in larger and smaller cities as well as in rural areas.

All  $\text{O}_3$  concentrations were below the maximum daily 8 h health limit of  $120 \mu\text{g m}^{-3}$  (Fig. 4). The concentrations of CO were far away from both the 1 h and maximum daily 8 h health limits of  $10$  and  $5 \text{mg m}^{-3}$ , respectively (Fig. S7), and the situation was similar for  $\text{SO}_2$ , for which the 1 h and daily limits are  $250$  and  $125 \mu\text{g m}^{-3}$ , respectively (Fig. S8).

#### 4 Conclusions

The relationships between urban air quality and motor vehicle road traffic are not straightforward since the contributions of traffic flow to pollutant concentrations are superimposed in the variability in local meteorological conditions, long-range transport of air masses and other sources/sinks. We introduced here an approach based on both relative difference and standardised anomaly, which helps unfold some important confounding environmental factors. It can support the creation of a generalised picture on urban atmospheres.

The method was deployed on the Budapest data during the different phases of the first COVID-19 outbreak. Various restriction measures introduced due to the pandemic resulted in a decline of vehicle road traffic down to approximately 50 % during the severest limitations. In parallel, concentrations of  $\text{NO}$ ,  $\text{NO}_2$ , CO,  $N_{6-1000}$  and  $N_{6-100}$  decreased substantially, some other species such as  $\text{PM}_{2.5}$  mass,  $\text{PM}_{10}$  mass and  $N_{100-1000}$  changed modestly and inconclusively, while  $\text{O}_3$  showed an increasing tendency. Change rates of  $\text{NO}$  and  $\text{NO}_2$  with relative change in traffic intensity were the largest (approximately 0.6), total particle number concentration showed

considerable dependency (0.4), while variation of CO was modest (0.2). It was demonstrated that a similar decrease in concentrations as observed in the strictest pandemic phase can also be caused by other (natural/meteorological) effects than traffic. The rainy weather in June 2020 (the so-called St. Medard's 40 d of rain in central European folklore) yielded, for instance, very similar low pollution levels.

The study revealed that intentional reduction of traffic intensity can have unambiguous potentials for improving urban air quality as far as NO, NO<sub>2</sub>, CO and particle number concentrations are concerned. It should be added that the most critical pollutant in many European cities, including Budapest, namely the PM<sub>10</sub> mass, however, did not seem to be considerably affected by vehicle flow. Nevertheless, measures for tranquillising urban traffic can contribute to improved air quality through a new strategy for lowering the population exposure of inhabitants instead of high-risk management of individuals.

The method could be expanded by other important ordinary chemical species such as soot and by other location types such as near-city or regional background sites jointly with central locations in order to obtain more exact meteorology-normalised changes. The results also point to the importance of non-linear relationships among precursors and secondary pollutants, which are to be further studied to gain better insights into urban atmospheric chemistry and air quality issues.

*Data availability.* The observational data are accessible at <http://www.levegominoseg.hu/> (last access: 25 September 2020) (Hungarian Meteorological Service, 2020) or are available from the corresponding author – except for the vehicle road traffic – upon request.

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*Author contributions.* IS conceived the study. AZG, WT, TW and IS performed most aerosol and meteorological measurements. All the co-authors participated in the data processing, calculations and interpretation of the results. The figures were created by MV and AZG. IS wrote the manuscript with comments from all the coauthors.

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

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## References

- C3S (Copernicus Climate Change Service): ERA5: Fifth generation of ECMWF atmospheric reanalyses of the global climate, Copernicus Climate Change Service Climate Data Store, available at: <http://cds.climate.copernicus.eu> (last access: 1 August 2020), 2017.
- CAMS (Copernicus Atmosphere Monitoring Service): User Guide: Regional Production, Updated documentation covering all Regional operational systems and the ENSEMBLE, Report issued by Météo-France, available at: [https://atmosphere.copernicus.eu/sites/default/files/2020-09/CAMS50\\_2018SC2\\_D2.0.2-U2\\_Models\\_documentation\\_202003\\_v2.pdf](https://atmosphere.copernicus.eu/sites/default/files/2020-09/CAMS50_2018SC2_D2.0.2-U2_Models_documentation_202003_v2.pdf) (last access: 27 August 2020), 2019.
- Conticini, E., Frediani, B., and Caro, D.: Can atmospheric pollution be considered a co-factor in extremely high level of SARS-CoV-2 lethality in Northern Italy?, *Environ. Pollut.*, 261, 114465, <https://doi.org/10.1016/j.envpol.2020.114465>, 2020.
- de Jesus, A. L., Rahman, M. M., Mazaheri, M., Thompson, H., Knibbs, L. D., Jeong, C., Evans, G., Nei, W., Ding, A., Qiao, L., Li, L., Portin, H., Niemi, J. V., Timonen, H., Luoma, K., Petäjä, T., Kulmala, M., Kowalski, M., Peters, A., Cyrys, J., Ferrero, L., Manigrasso, M., Avino, P., Buonano, G., Reche, C., Querol, X., Beddows, D., Harrison, R. M., Sowlat, M. H., Sioutas, C., and Morawska, L.: Ultrafine particles and PM<sub>2.5</sub> in the air of cities around the world: Are they representative of each other?, *Environ. Int.*, 129, 118–135, 2019.
- EU Directives: Directive 2008/50/EC of the European Parliament and of the Council of 21 May 2008 on ambient air quality and cleaner air for Europe, Off. J. EU, L 152, Brussels, Belgium, 11 June 2008, 44 pp., 2008.
- Frontera, A., Cianfanelli, L., Vlachos, K., Landoni, G., and Cremona, G.: Severe air pollution links to higher mortality in COVID-19 patients: The “double-hit” hypothesis, *J. Infection*, 81, 255–259, <https://doi.org/10.1016/j.jinf.2020.05.031>, 2020.
- Gentner, D. R., Jathar, S. H., Gordon, T. D., Bahreini, R., Day, D. A., El Haddad, I., Hayes, P. L., Pieber, S. M., Platt, S. M., de Gouw, J., Goldstein, A. H., Harley, R. A., Jimenez, J. L., Prévôt, A. S. H., and Robinson, A. L.: Review of urban secondary organic aerosol formation from gasoline and diesel motor vehicle emissions, *Environ. Sci. Technol.*, 51, 1074–1093, 2017.

- Harrison, R. M.: Urban atmospheric chemistry: a very special case for study, *Clim. Atmos. Sci.*, 1, 20175, <https://doi.org/10.1038/s41612-017-0010-8>, 2018.
- Harrison, R. M., Jones, A. M., Gietl, J., Yin, J., and Green, D. C.: Estimation of the contributions of brake dust, tire wear, and resuspension to nonexhaust traffic particles derived from atmospheric measurements, *Environ. Sci. Technol.*, 46, 6523–6529, 2012.
- Hopke, P. K.: Review of receptor modeling methods for source apportionment, *J. Air Waste Manage.*, 66, 237–259, 2016.
- Horvath, H., Kreiner, I., Norek, C., Preining O., and Georgi, B.: Diesel emissions in Vienna, *Atmos. Environ.*, 22, 1255–1269, 1988.
- Hungarian Meteorological Service: Homepage, Hungarian Air Quality Monitoring Network, available at: <http://www.levegominoseg.hu/>, last access: 25 September 2020.
- Jacob, J. J.: Introduction to Atmospheric Chemistry, Princeton University Press, Cambridge, USA, 1999.
- 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., Ryan, R. G., Fakes, L. H., and Pawson, S.: Global Impact of COVID-19 Restrictions on the Surface Concentrations of Nitrogen Dioxide and Ozone, *Atmos. Chem. Phys. Discuss.*, <https://doi.org/10.5194/acp-2020-685>, in review, 2020.
- Lal, P., Kumar, A., Kumar, S., Kumari, S., Saikia, P., Dayanandan, A., Adhikari, D., and Khane, M. L.: The dark cloud with a silver lining: Assessing the impact of the SARS COVID-19 pandemic on the global environment, *Sci. Total Environ.*, 732, 139297, <https://doi.org/10.1016/j.scitotenv.2020.139297>, 2020.
- Le, T., Wang, Y., Liu, L., Yang, J., Yung, Y. L., Li, G., and Seinfeld, J. H.: Unexpected air pollution with marked emission reductions during the COVID-19 outbreak in China, *Science*, 369, 702–706, 2020.
- Lee, J. D., Drysdale, W. S., Finch, D. P., Wilde, S. E., and Palmer, P. I.: UK surface NO<sub>2</sub> levels dropped by 42 % during the COVID-19 lockdown: impact on surface O<sub>3</sub>, *Atmos. Chem. Phys. Discuss.*, <https://doi.org/10.5194/acp-2020-838>, in review, 2020.
- Lelieveld, J. and Dentener, F. J.: What controls tropospheric ozone?, *J. Geophys. Res.-Atmos.*, 105, 3531–3551, 2000.
- Liu, Y., Ning, Z., Chen, Y., Guo, M., Liu, Y., Gali, N. K., Sun, L., Duan, Y., Cai, J., Westerdahl, D., Liu, X., Xu, K., Ho, K., Kan, H., Fu, Q., and Lan, K.: Aerodynamic analysis of SARS-CoV-2 in two Wuhan hospitals, *Nature*, 582, 557–560, <https://doi.org/10.1038/s41586-020-2271-3>, 2020.
- Mahato, S., Pal, S., and Ghosh, K. G.: Effect of lockdown amid COVID-19 pandemic on air quality of the megacity Delhi, India, *Sci. Total Environ.*, 730, 39086, <https://doi.org/10.1016/j.scitotenv.2020.139086>, 2020.
- Maheras, P., Tolika, K., Tegoulas, I., Anagnostopoulou, Ch., Szpirosz, K., Károssy, Cs., and Makra, L.: Comparison of an automated classification system with an empirical classification of circulation patterns over the Pannonian basin, Central Europe, *Meteorol. Atmos. Phys.*, 131, 739–751, <https://doi.org/10.1007/s00703-018-0601-x>, 2018.
- Marécal, V., Peuch, V.-H., Andersson, C., Andersson, S., Arteta, J., Beekmann, M., Benedictow, A., Bergström, R., Bessagnet, B., Cansado, A., Chéroux, F., Colette, A., Coman, A., Curier, R. L., Denier van der Gon, H. A. C., Drouin, A., Elbern, H., Emili, E., Engelen, R. J., Eskes, H. J., Foret, G., Friese, E., Gauss, M., Giannaros, C., Guth, J., Joly, M., Jaumouillé, E., Josse, B., Kadyrov, N., Kaiser, J. W., Krajsek, K., Kuenen, J., Kumar, U., Liora, N., Lopez, E., Malherbe, L., Martinez, I., Melas, D., Meleux, F., Menut, L., Moinat, P., Morales, T., Parmentier, J., Piacentini, A., Plu, M., Poupkou, A., Queguiner, S., Robertson, L., Rouïl, L., Schaap, M., Segers, A., Sofiev, M., Tarasson, L., Thomas, M., Timmermans, R., Valdebenito, Á., van Velthoven, P., van Versendaal, R., Vira, J., and Ung, A.: A regional air quality forecasting system over Europe: the MACC-II daily ensemble production, *Geosci. Model Dev.*, 8, 2777–2813, <https://doi.org/10.5194/gmd-8-2777-2015>, 2015.
- Mikkonen, S., Németh, Z., Varga, V., Weidinger, T., Leinonen, V., Yli-Juuti, T., and Salma, I.: Decennial time trends and diurnal patterns of particle number concentrations in a central European city between 2008 and 2018, *Atmos. Chem. Phys.*, 20, 12247–12263, <https://doi.org/10.5194/acp-20-12247-2020>, 2020.
- Morawska, L. and Cao, J.: Airborne transmission of SARS-CoV-2: The world should face the reality, *Environ. Int.*, 139, 105730, <https://doi.org/10.1016/j.envint.2020.105730>, 2020.
- Nakada, L. Y. K. and Urban, R. C.: COVID-19 pandemic: Impacts on the air quality during the partial lockdown in São Paulo state, Brazil, *Sci. Total Environ.*, 730, 139087, <https://doi.org/10.1016/j.scitotenv.2020.139087>, 2020.
- Paasonen, P., Kupiainen, K., Klimont, Z., Visschedijk, A., Denier van der Gon, H. A. C., and Amann, M.: Continental anthropogenic primary particle number emissions, *Atmos. Chem. Phys.*, 16, 6823–6840, <https://doi.org/10.5194/acp-16-6823-2016>, 2016.
- Péczei, Gy.: Grosswetterlagen in Ungarn (Large-scale weather situations in Hungary, Publication of the Hungarian Meteorological Institute, 30, 86 pp., Budapest, Hungary, 1957 (in German).
- 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<sub>2</sub> pollution in Spain, *Atmos. Chem. Phys.*, 20, 11119–11141, <https://doi.org/10.5194/acp-20-11119-2020>, 2020.
- Putaud, J.-P., Van Dingenen, R., Alastuey, A., Bauer, H., Birmili, W., Cyrys, J., Flentje, H., Fuzzi, S., Gehrig, R., Hansson, H. C., Harrison, R. M., Herrmann, H., Hitznerberger, R., Hüglin, C., Jones, A. M., Kasper-Giebl, A., Kiss, G., Kousa, A., Kuhlbusch, T. A. J., Löschau, G., Maenhaut, W., Molnár, A., Moreno, T., Pekkanen, J., Perrino, C., Pitz, M., Puxbaum, H., Querol, X., Rodriguez, S., Salma, I., Schwarz, J., Smolík, J., Schneider, J., Spindler, G., ten Brink, H., Turšič, J., Viana, M., Wiedensohler, A., and Raes, F.: A European Aerosol Phenomenology – 3: physical and chemical characteristics of particulate matter from 60 rural, urban, and kerbside sites across Europe, *Atmos. Environ.*, 44, 1308–1320, 2010.
- Rönkkö, T., Kuuluvainen, H., Karjalainen, P., Keskinen, J., Hillamo, R., Niemi, J. V., Pirjola, L., Timonen, H. J., Saarikoski, S., Saukko, E., Järvinen, A., Silvennoinen, H., Rostedt, A., Olin, M., Yli-Ojanperä, J., Nousiainen, P., Kousa, A., and Dal Maso, M.: Traffic is a major source of atmospheric nanocluster aerosol, *P. Natl. Acad. Sci. USA*, 114, 7549–7554, 2017.
- Salma, I. and Maenhaut, W.: Changes in chemical composition and mass of atmospheric aerosol pollution between 1996 and 2002 in a Central European city, *Environ. Pollut.*, 143, 479–488, 2006.

- Salma, I. and Németh, Z.: Dynamic and timing properties of new aerosol particle formation and consecutive growth events, *Atmos. Chem. Phys.*, 19, 5835–5852, <https://doi.org/10.5194/acp-19-5835-2019>, 2019.
- Salma, I., Borsós, T., Németh, Z., Weidinger, T., Aalto, T., and Kulmala, M.: Comparative study of ultrafine atmospheric aerosol within a city, *Atmos. Environ.*, 92, 154–161, 2014.
- Salma, I., Németh, Z., Weidinger, T., Kovács, B., and Kristóf, G.: Measurement, growth types and shrinkage of newly formed aerosol particles at an urban research platform, *Atmos. Chem. Phys.*, 16, 7837–7851, <https://doi.org/10.5194/acp-16-7837-2016>, 2016a.
- Salma, I., Németh, Z., Kerminen, V.-M., Aalto, P., Nieminen, T., Weidinger, T., Molnár, Á., Imre, K., and Kulmala, M.: Regional effect on urban atmospheric nucleation, *Atmos. Chem. Phys.*, 16, 8715–8728, <https://doi.org/10.5194/acp-16-8715-2016>, 2016b.
- Salma, I., Varga, V., and Németh, Z.: Quantification of an atmospheric nucleation and growth process as a single source of aerosol particles in a city, *Atmos. Chem. Phys.*, 17, 15007–15017, <https://doi.org/10.5194/acp-17-15007-2017>, 2017.
- Salma, I., Vasánits-Zsigrai, A., Machon, A., Varga, T., Major, I., Gergely, V., and Molnár, M.: Fossil fuel combustion, biomass burning and biogenic sources of fine carbonaceous aerosol in the Carpathian Basin, *Atmos. Chem. Phys.*, 20, 4295–4312, <https://doi.org/10.5194/acp-20-4295-2020>, 2020a.
- Salma, I., Thén, W., Aalto, P., Kerminen, V.-M., Kern, A., Barcza, Z., Petäjä, T., and Kulmala, M.: Influence of vegetation on occurrence and time distributions of regional new aerosol particle formation and growth, *Atmos. Chem. Phys. Discuss.*, <https://doi.org/10.5194/acp-2020-862>, in review, 2020b.
- Shaman, J. and Kohn, M.: Absolute humidity modulates influenza survival, transmission, and seasonality, *P. Natl. Acad. Sci. USA*, 106, 3243–3248, 2009.
- Sussmann, R. and Rettinger, M.: Can we measure a COVID-19-related slowdown in atmospheric CO<sub>2</sub> growth? Sensitivity of total carbon column observations, *Remote Sens.*, 12, 2387, <https://doi.org/10.3390/rs12152387>, 2020.
- Tobías, A., Carnerero, C., Reche, C., Massagué, J., Via, M., Minguillón, M. C., Alastuey, A., and Querol, X.: Changes in air quality during the lockdown in Barcelona (Spain) one month into the SARS-CoV-2 epidemic, *Sci. Total Environ.*, 726, 138540, <https://doi.org/10.1016/j.scitotenv.2020.138540>, 2020.
- VM 4: A levegőterheltségi szint határértékeiről és a helyhez kötött légszennyezőpontforrások kibocsátási határértékeiről (On the limit values of ambient air quality and emissions from fixed sources, *Magyar Közlöny* 4, 487–533, 2011 (in Hungarian).
- Wang, P., Chen, K., Zhu, S., Wang, P., and Zhang, H.: Severe air pollution events not avoided by reduced anthropogenic activities during COVID-19 outbreak, *Resour. Conserv. Recycl.*, 158, 104814, <https://doi.org/10.1016/j.resconrec.2020.104814>, 2020.
- Warneck, P. and Williams, J.: *The Atmospheric Chemist's Companion, Numerical Data for Use in the Atmospheric Sciences*, Springer, Dordrecht, the Netherlands, 2012.
- WHO (World Health Organization): Coronavirus disease 2019 (COVID-19): situation report, 51. World Health Organization, available at: <https://apps.who.int/iris/handle/10665/331475>, last access: 9 August 2020.
- WMO (World Meteorological Organization): *Guide to Meteorological Instruments and Methods of Observation*, No. 8, Appendix 4B, Geneva, Switzerland, 2008.

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*Supplement of*

**What can we learn about urban air quality with regard to  
the first outbreak of the COVID-19 pandemic?  
A case study from central Europe**

**Imre Salma et al.**

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**Table S1.** Macrocirculation patterns and their seasonal and annual occurrences in the Carpathian Basin for years 1958–2010 (Maheras et al., 2018).

No.	Code	Description	Occurrence frequency (%)				
			Winter	Spring	Summer	Autumn	Annual
1	mCc	Cyclone with a cold front over northeastern Europe, northerly wind	7.3	11.3	12.1	8.0	9.7
2	AB	Anticyclone over the British Isles, northerly wind	5.6	7.1	8.6	6.4	6.9
3	CMc	Mediterranean cyclone with a cold front over southern Europe, northerly wind	2.5	3.5	1.8	1.9	2.4
4	mCw	Mediterranean cyclone with a warm front over northeastern Europe, southerly wind	9.2	9.7	5.7	7.2	7.9
5	Ae	Anticyclone over eastern Europe, southerly wind	14.2	11.3	7.3	17.6	12.6
6	CMw	Mediterranean cyclone with a warm front over southern Europe, southerly wind	8.9	8.7	3.7	8.3	7.4
7	zC	Highly developed cyclone over northern Europe, westerly wind	5.0	3.2	2.7	2.9	3.5
8	Aw	Anticyclone over western Europe, westerly wind	13.1	11.2	20.8	12.8	14.6
9	As	Anticyclone over southern Europe, westerly wind	7.0	4.4	2.9	5.6	4.9
10	An	Anticyclone over northern Europe, easterly wind	10.9	12.8	11.3	10.1	11.3
11	AF	Anticyclone over Fennoscandia, easterly wind	2.8	5.2	5.9	3.7	4.4
12	A	Anticyclone over the Carpathian Basin, changing wind direction	11.8	7.3	13.3	13.3	11.4
13	C	Cyclone over the Carpathian Basin, changing wind direction	1.7	4.3	3.9	2.2	3.0

Pandemic phase	Date (dd-MM)	Day of week	Heating info	MCP code	Precipitation	Pandemic phase	Date (dd-MM)	Day of week	Heating info	MCP code	Precipitation	Pandemic phase	Date (dd-MM)	Day of week	Heating info	MCP code	Precipitation
Pre-emergency (71 d)	01-01	Wed	Heating season (106 days)	8	0	Restriction (51 d)	12-03	Thu	Non-heating season (107 days)	7	0	Post-emergency (44 d)	18-05	Mon	Non-heating season (cont'd)	8	0
	02-01	Thu		9	0		13-03	Fri		7	0		19-05	Tue		6	0
	03-01	Fri		12	0		14-03	Sat		8	0		20-05	Wed		1	1
	04-01	Sat		1	1		15-03	Sun		12	0		21-05	Thu		11	0
	05-01	Sun		8	0		16-03	Mon		5	0		22-05	Fri		11	0
	06-01	Mon		12	0		17-03	Tue		5	0		23-05	Sat		4	1
	07-01	Tue		12	0		18-03	Wed		12	0		24-05	Sun		4	0
	08-01	Wed		12	0		19-03	Thu		12	0		25-05	Mon		2	1
	09-01	Thu		12	0		20-03	Fri		12	0		26-05	Tue		2	0
	10-01	Fri		9	0		21-03	Sat		5	1		27-05	Wed		2	3
	11-01	Sat		8	0		22-03	Sun		2	0		28-05	Thu		2	1
	12-01	Sun		12	0		23-03	Mon		10	1		29-05	Fri		11	1
	13-01	Mon		12	0		24-03	Tue		10	0		30-05	Sat		11	1
	14-01	Tue		12	0		25-03	Wed		10	0		31-05	Sun		11	2
	15-01	Wed		5	0		26-03	Thu		10	0		01-06	Mon		11	0
	16-01	Thu		5	0		27-03	Fri		10	0		02-06	Tue		11	0
	17-01	Fri		12	0		28-03	Sat		5	0		03-06	Wed		4	0
	18-01	Sat		5	0		29-03	Sun		2	0		04-06	Thu		1	0
	19-01	Sun		2	0		30-03	Mon		2	0		05-06	Fri		1	5
	20-01	Mon		2	0		31-03	Tue		2	0		06-06	Sat		1	0
	21-01	Tue		2	0		01-04	Wed		2	0		07-06	Sun		1	11
	22-01	Wed		2	0		02-04	Thu		12	0		08-06	Mon		1	0
	23-01	Thu		2	0		03-04	Fri		12	0		09-06	Tue		1	7
	24-01	Fri		9	0		04-04	Sat		10	0		10-06	Wed		13	4
	25-01	Sat		9	0		05-04	Sun		10	0		11-06	Thu		13	0
	26-01	Sun		12	0		06-04	Mon		5	0		12-06	Fri		13	0
	27-01	Mon		9	0		07-04	Tue		5	0		13-06	Sat		4	2
	28-01	Tue		1	8		08-04	Wed		12	0		14-06	Sun		4	24
	29-01	Wed		1	1		09-04	Thu		12	0		15-06	Mon		1	26
	30-01	Thu		13	0		10-04	Fri		2	0		16-06	Tue		4	4
	31-01	Fri		4	1		11-04	Sat		10	0		17-06	Wed		4	13
01-02	Sat	7	1	12-04	Sun	5	0	18-06	Thu	4	5						
02-02	Sun	7	2	13-04	Mon	5	0	19-06	Fri	1	0						
03-02	Mon	7	1	14-04	Tue	1	0	20-06	Sat	8	0						
04-02	Tue	13	6	15-04	Wed	2	0	21-06	Sun	8	1						
05-02	Wed	6	0	16-04	Thu	9	0	22-06	Mon	8	0						
06-02	Thu	2	0	17-04	Fri	9	0	23-06	Tue	8	0						
07-02	Fri	10	0	18-04	Sat	11	0	24-06	Wed	11	0						
08-02	Sat	12	0	19-04	Sun	13	0	25-06	Thu	11	0						
09-02	Sun	9	0	20-04	Mon	11	0	26-06	Fri	10	1						
10-02	Mon	1	5	21-04	Tue	11	0	27-06	Sat	5	0						
11-02	Tue	13	0	22-04	Wed	11	0	28-06	Sun	12	0						
12-02	Wed	8	0	23-04	Thu	11	0	29-06	Mon	1	6						
13-02	Thu	9	2	24-04	Fri	12	0	30-06	Tue	8	0						
14-02	Fri	1	0	25-04	Sat	1	0	01-07	Wed	9	0						
15-02	Sat	12	0	26-04	Sun	1	0	02-07	Thu	1	2						
16-02	Sun	5	0	27-04	Mon	1	0	03-07	Fri	1	0						
17-02	Mon	5	0	28-04	Tue	1	0	04-07	Sat	8	0						
18-02	Tue	7	0	29-04	Wed	1	3	05-07	Sun	8	0						
19-02	Wed	4	1	30-04	Thu	4	4	06-07	Mon	8	7						
20-02	Thu	4	0	01-05	Fri	1	1	07-07	Tue	8	0						
21-02	Fri	5	0	02-05	Sat	1	0	08-07	Wed	8	0						
22-02	Sat	8	0	03-05	Sun	1	0	09-07	Thu	12	0						
23-02	Sun	7	0	04-05	Mon	4	0	10-07	Fri	12	0						
24-02	Mon	1	0	05-05	Tue	13	0	11-07	Sat	1	7						
25-02	Tue	4	2	06-05	Wed	1	0	12-07	Sun	8	0						
26-02	Wed	1	6	07-05	Thu	2	0	13-07	Mon	8	0						
27-02	Thu	1	1	08-05	Fri	12	0	14-07	Tue	12	0						
28-02	Fri	13	2	09-05	Sat	9	0	15-07	Wed	12	0						
29-02	Sat	9	0	10-05	Sun	1	0	16-07	Thu	1	0						
01-03	Sun	1	17	11-05	Mon	6	0	17-07	Fri	1	2						
02-03	Mon	13	2	12-05	Tue	1	1	18-07	Sat	1	1						
03-03	Tue	13	4	13-05	Wed	1	0	19-07	Sun	10	7						
04-03	Wed	13	0	14-05	Thu	13	0	20-07	Mon	1	0						
05-03	Thu	12	2	15-05	Fri	13	0	21-07	Tue	1	1						
06-03	Fri	13	11	16-05	Sat	2	2	22-07	Wed	1	0						
07-03	Sat	13	0	17-05	Sun	2	3	23-07	Thu	1	1						
08-03	Sun	8	0					24-07	Fri	10	17						
09-03	Mon	6	0					25-07	Sat	13	0						
10-03	Tue	6	1					26-07	Sun	1	2						
11-03	Wed	7	0					27-07	Mon	1	0						
								28-07	Tue	1	0						
								29-07	Wed	1	0						
								30-07	Thu	8	0						
								31-07	Fri	12	0						

Total number of days: 213

**Figure S1.** Survey on time intervals of interest with basic facts, daily macrocirculation pattern (MCP) codes and daily precipitation sum (mm) in Budapest from 1 January to 31 July 2020. The holidays are indicated in darker green, heating season in darker grey, non-heating season in lighter grey, Pre-emergency phase of the first COVID-19 outbreak in lighter blue, Pre-restriction phase in lighter yellow, Restriction phase in orange, Post-restriction phase in darker yellow and Post-emergency phase in darker blue colours. The cyclonic and anticyclonic MCP types are marked in lighter green and pink colours, respectively.

Motor vehicles road traffic was measured by the Budapest Public Roads Ltd., which is responsible for operation, control and maintenance of all roads, streets, bridges, tunnels, other structures and traffic engineering facilities in Budapest. The actual locations were 1) Szabadság Bridge, 2) Váci Road near its junction with Árpád Road, 3) Alkotás Road near its junction with Nagyenyed Street, 4) Margit Boulevard and Vérmező Street which both lead to Széna Square (and, therefore, this site is called here as Széna Square), where the air quality monitoring station is also located.

The Szabadság Bridge hosts quite complex events. The bridge was reserved for the pedestrians and was closed for motor vehicles on some summer weekends, namely on 17–18 June, 24–25 June, 5–6 August and 12–13 August in 2017, on 14–15 July, 21–22 July, 28–29 July and 4–5 August in 2018, on 6–7 July, 13–14 July, 20–21 July and 27–28 July on 2019. It was also partially or completely closed for vehicles in some other intervals due to urban running races or for its planned extended cleaning. The vehicle census data for these days and time intervals were included in deriving the time series of vehicular traffic but were excluded when calculating the average diurnal patterns and descriptive statistics to avoid their distortion due to these very specific or unusual circumstances.

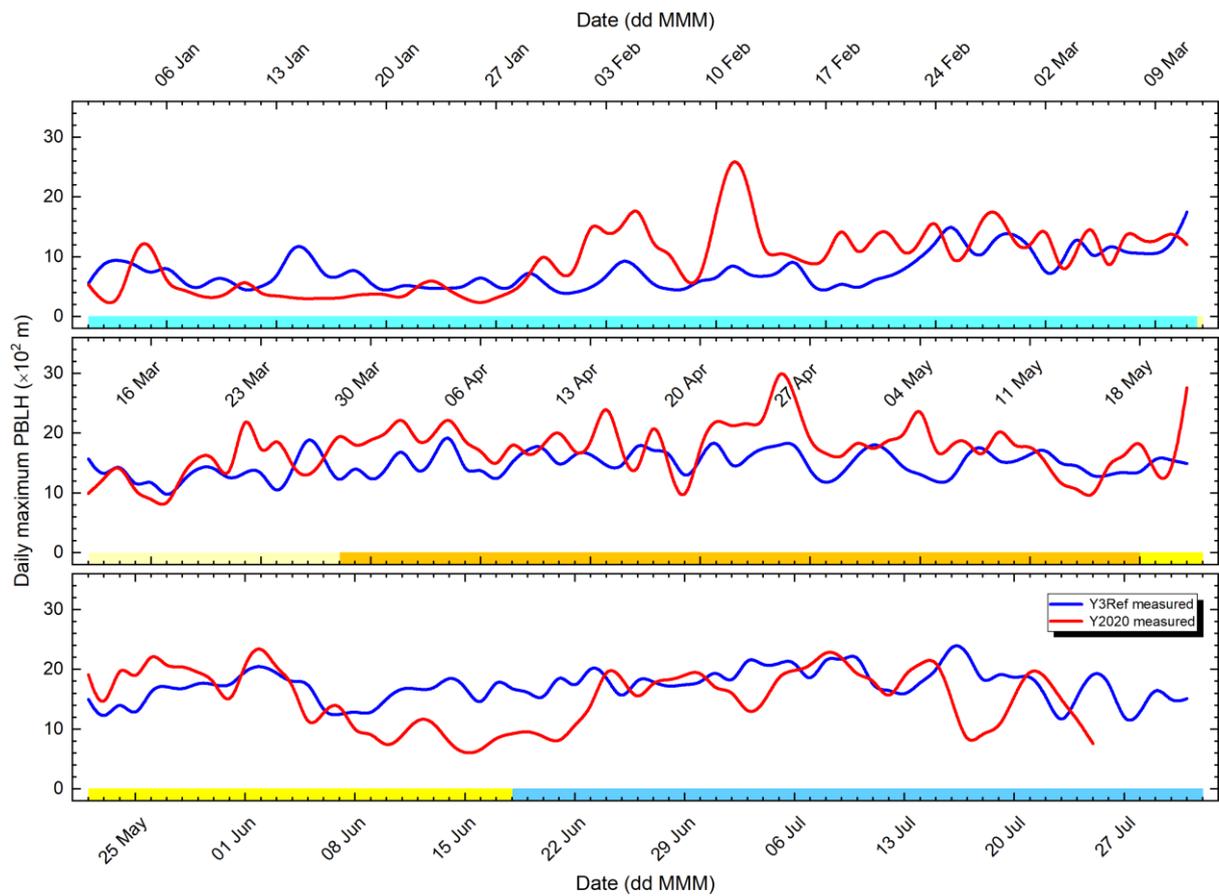
For location no. 3 (Alkotás Road), the data coverage for year 2019 was poor (21 %) and, therefore, this year was excluded from the averaging for the reference year. Similarly, there were larger scale traffic control arrangements and missing data in larger abundance for the site no. 4 (Széna Square), and, therefore, year 2019 was only maintained here.

**Table S2.** Ranges and medians of hourly mean  $T$  ( $^{\circ}\text{C}$ ), RH (%), absolute humidity (AH, in  $\text{g m}^{-3}$ ), WS ( $\text{m s}^{-1}$ ), GRad ( $\text{W m}^{-2}$ , for individual data  $\geq 50 \text{ W m}^{-2}$ ), daily maximum planetary boundary layer height (PBLH<sub>max</sub>, in km) in the average reference year of 2017–2019 (Y3Ref) and year 2020 (Y2020) together with their relative difference (RDiff) in % and their anomaly standardised to SD (SAly) for the overall state of emergency time interval (from 12 March to 17 June).

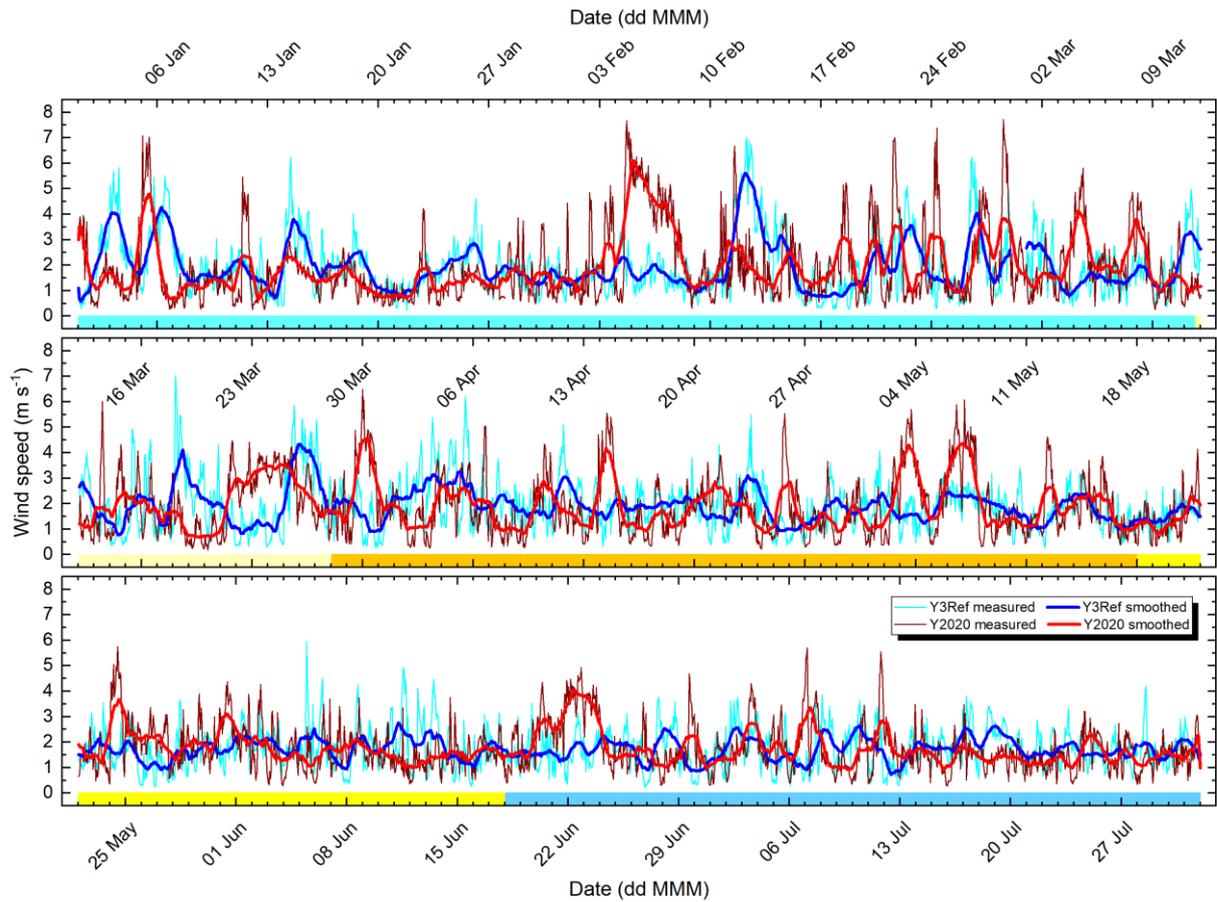
Interval/ Variable	Y3Ref			Y2020			RDiff	SAly
	Min	Median	Max	Min	Median	Max		
$T$	3.2	16	30	-0.4	15	31	-0.7 <sup>†</sup>	-0.1
RH	19	56	92	12	50	100	-11	-0.4
AH	2.8	7.6	17	1.2	5.9	17	-22	-0.5
WS	0.3	1.7	7.0	0.2	1.6	6.5	-6	-0.1
GRad	51	408	913	50	405	977	-1	-3 <sup>‡</sup>
PBLH <sub>max</sub>	0.84	1.5	2.2	0.57	1.7	3.3	+10	+0.3

<sup>†</sup> Y2020–Y3Ref difference in median  $T$ s; in a unit of  $^{\circ}\text{C}$ .

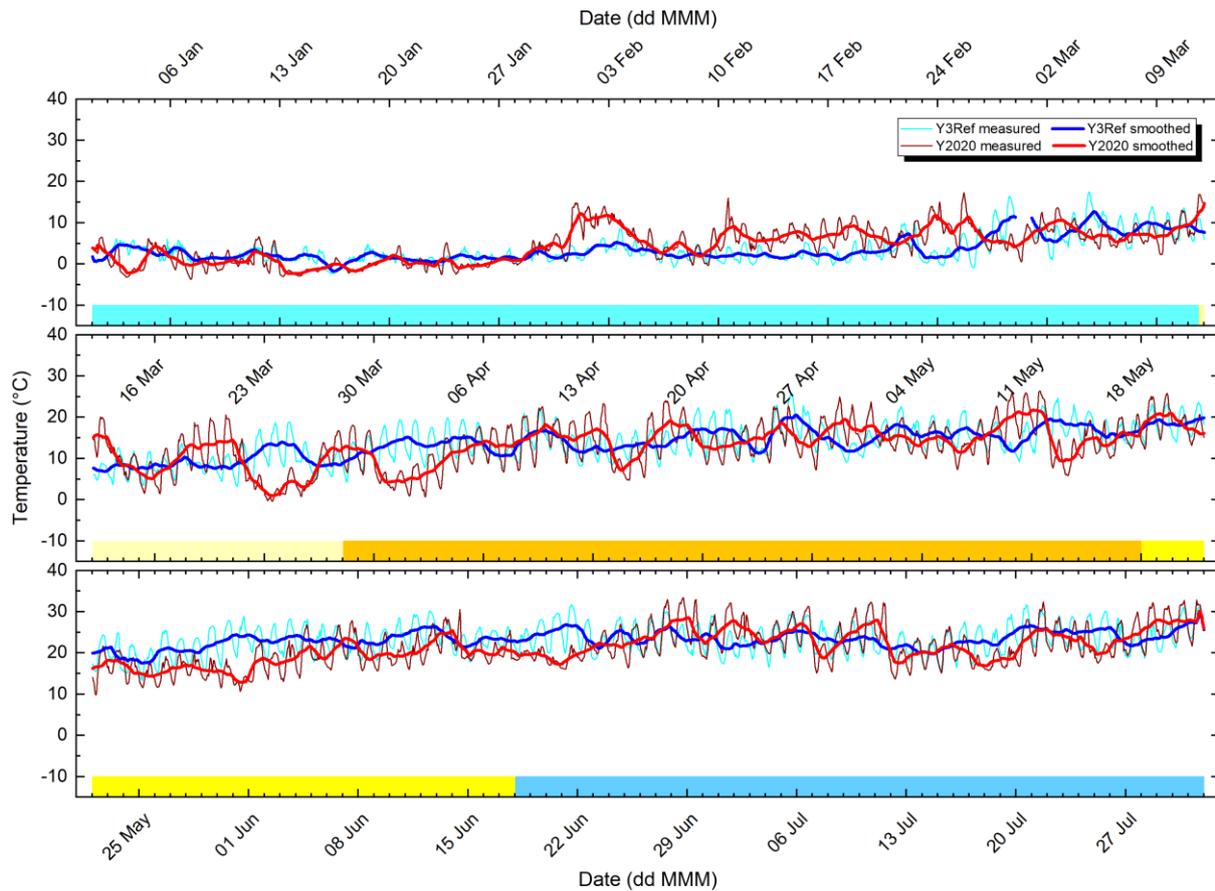
<sup>‡</sup> Anomaly not standardised to SD; in a unit of  $\text{W m}^{-2}$ .



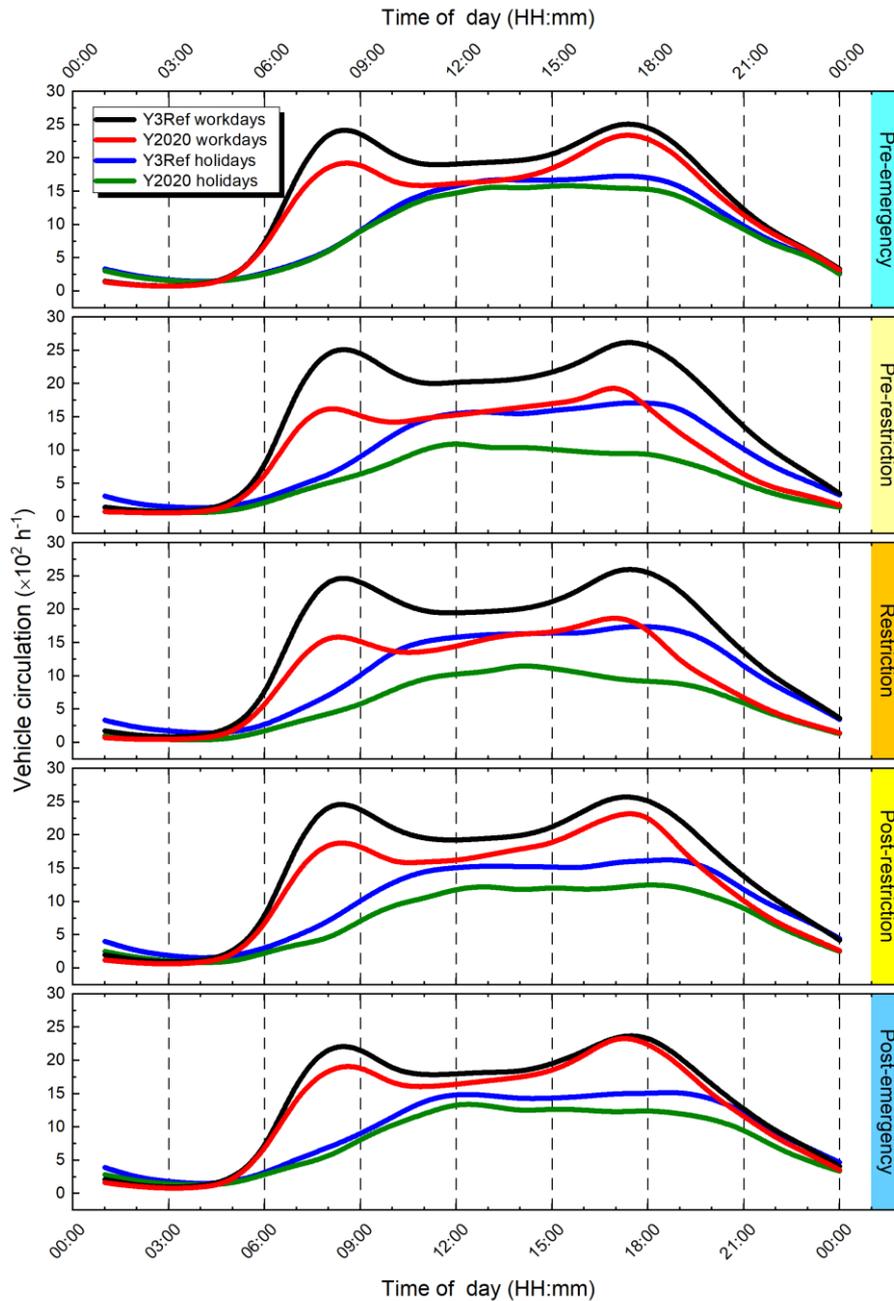
**Figure S2.** Time series of the daily maximum of planetary boundary layer height ( $PBLH_{max}$ ) in the average reference year of 2017–2019 (Y3Ref) and year 2020 over the five consecutive phases of the first COVID-19 outbreak. The phases are marked by the following colour codes: Pre-emergency phase lighter blue, Pre-restriction phase lighter yellow, Restriction phase orange, Post-restriction phase darker yellow and Post-emergency phase darker blue.



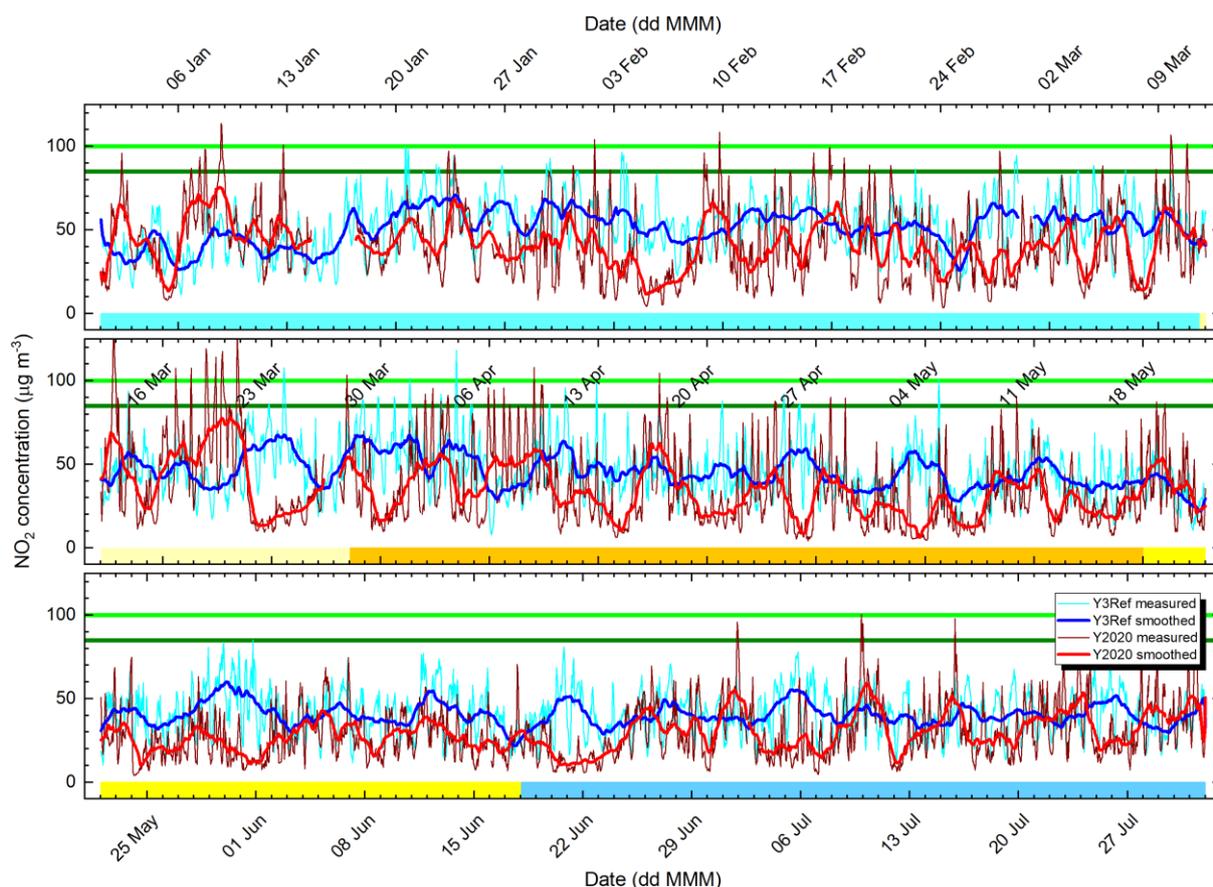
**Figure S3.** Time series of wind speed in the average reference year of 2017–2019 (Y3Ref) and year 2020 together with their 24-h smoothed curves over the five consecutive phases of the first COVID-19 outbreak. The phases are marked by the following colour codes: Pre-emergency phase lighter blue, Pre-restriction phase lighter yellow, Restriction phase orange, Post-restriction phase darker yellow and Post-emergency phase darker blue. The tick labels of the abscissa indicate the Mondays in 2020.



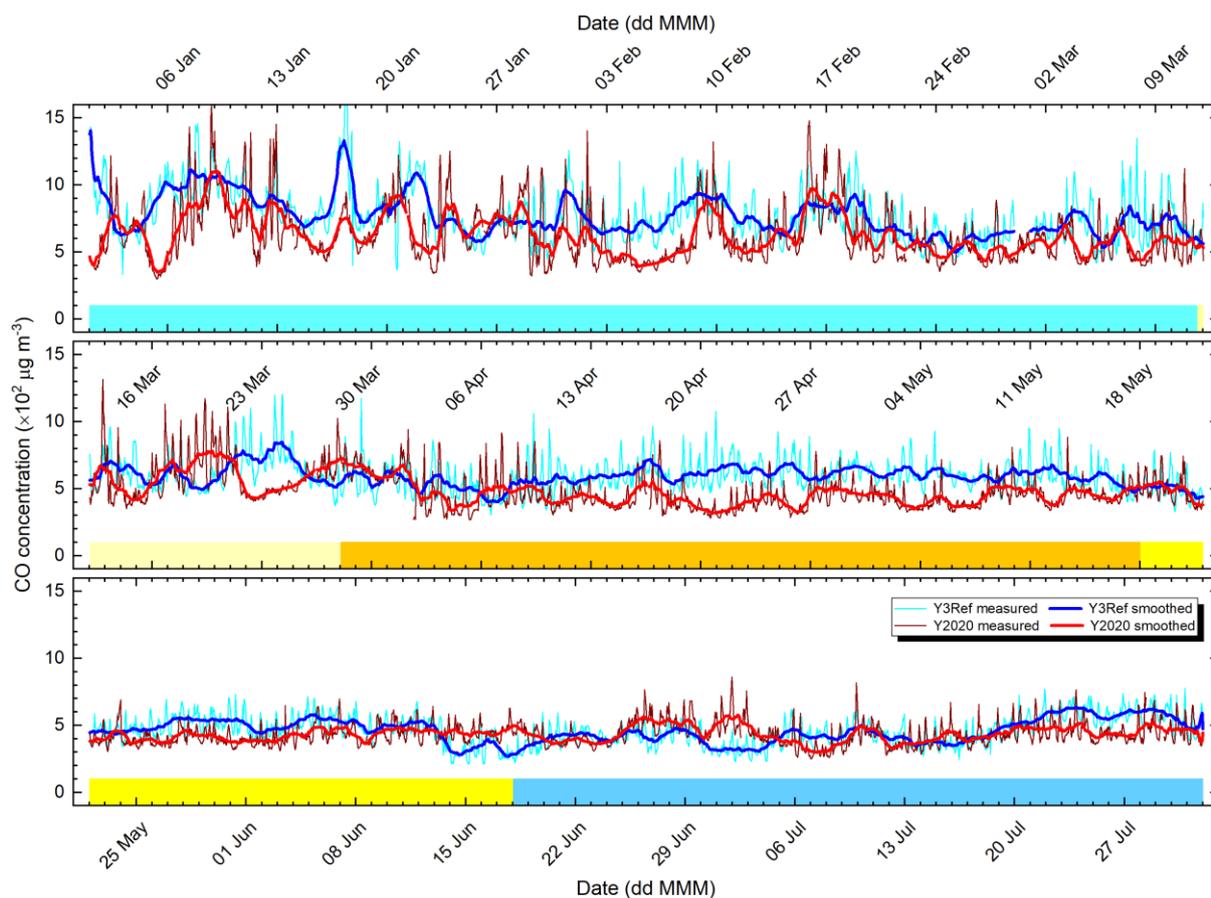
**Figure S4.** Time series of air temperature in the average reference year of 2017–2019 (Y3Ref) and year 2020 together with their 24-h smoothed curves over the five consecutive phases of the first COVID-19 outbreak. The phases are marked by the following colour codes: Pre-emergency phase lighter blue, Pre-restriction phase lighter yellow, Restriction phase orange, Post-restriction phase darker yellow and Post-emergency phase darker blue. The tick labels of the abscissa indicate the Mondays in 2020.



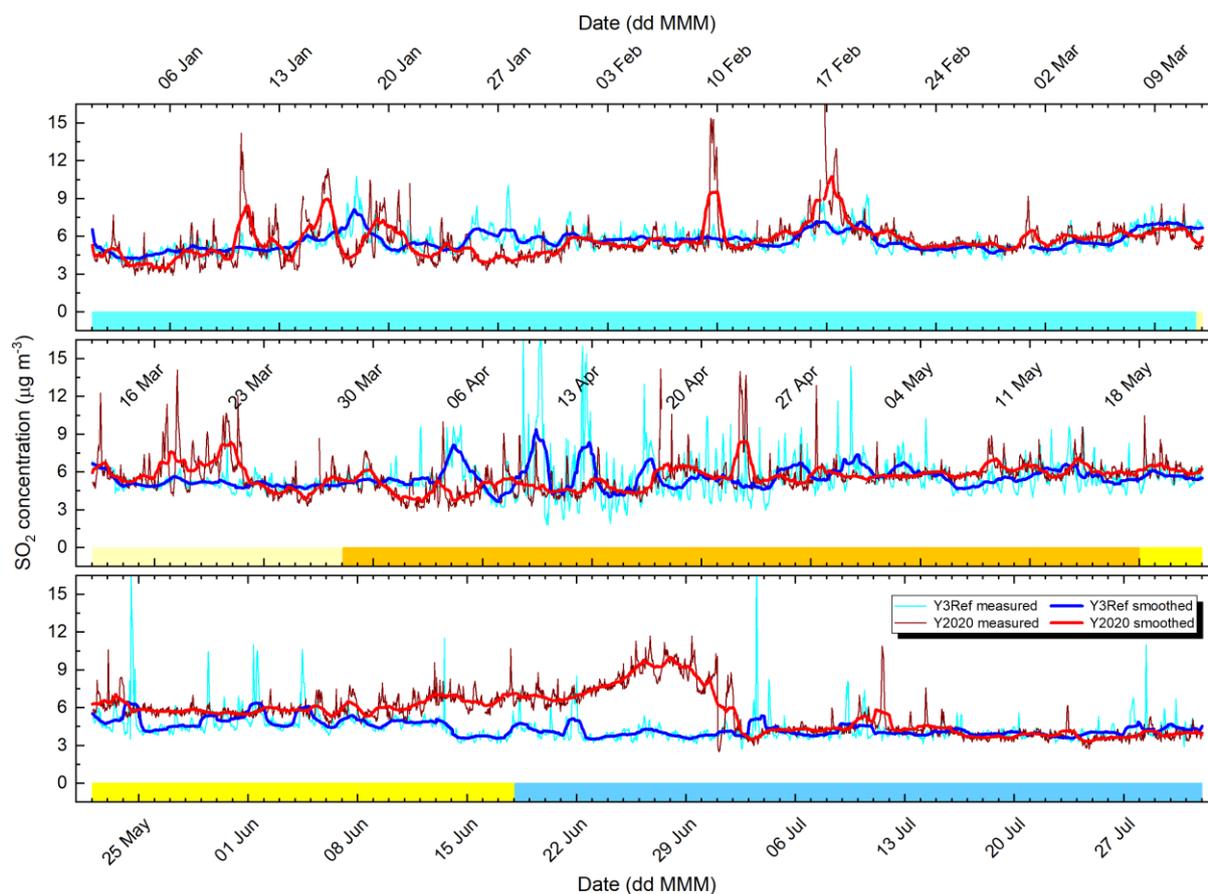
**Figure S5.** Average diurnal variations of motor vehicle road traffic in both directions on a major route (Váci Road) in Budapest separately for workdays and holidays in the average reference year of 2017–2019 (Y3Ref) and year 2020 over the five consecutive phases of the first COVID-19 outbreak. The phases are marked by the following colour codes: Pre-emergency phase lighter blue, Pre-restriction phase lighter yellow, Restriction phase orange, Post-restriction phase darker yellow and Post-emergency phase darker blue.



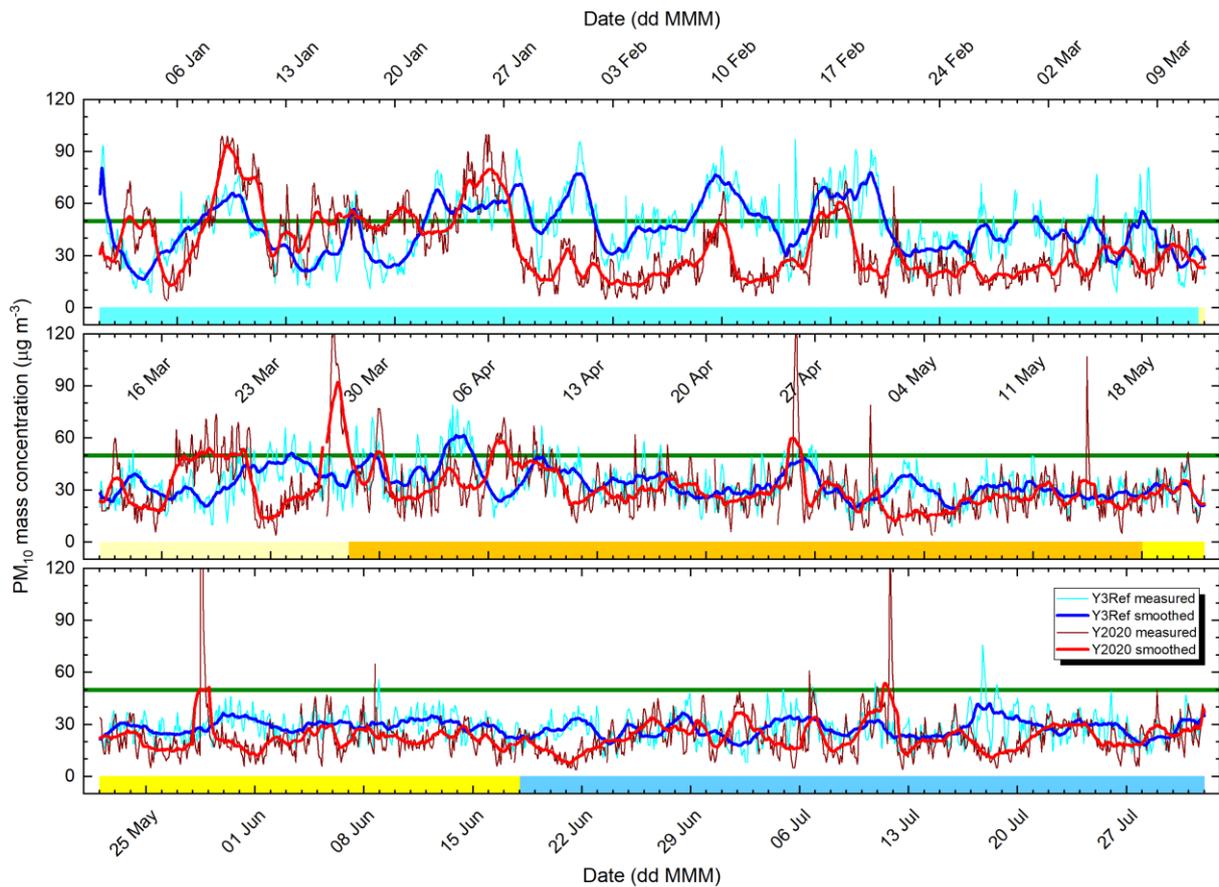
**Figure S6.** Time series of NO<sub>2</sub> concentration in the average reference year of 2017–2019 (Y3Ref) and year 2020 together with their 24-h smoothed curves over the five consecutive phases of the first COVID-19 outbreak. The phases are marked by the following colour codes: Pre-emergency phase lighter blue, Pre-restriction phase lighter yellow, Restriction phase orange, Post-restriction phase darker yellow and Post-emergency phase darker blue. The 1-h and daily health limits of 100 and 85  $\mu\text{g m}^{-3}$  are indicated by horizontal lines in lighter green and darker green colours, respectively. The tick labels of the abscissa indicate the Mondays in 2020.



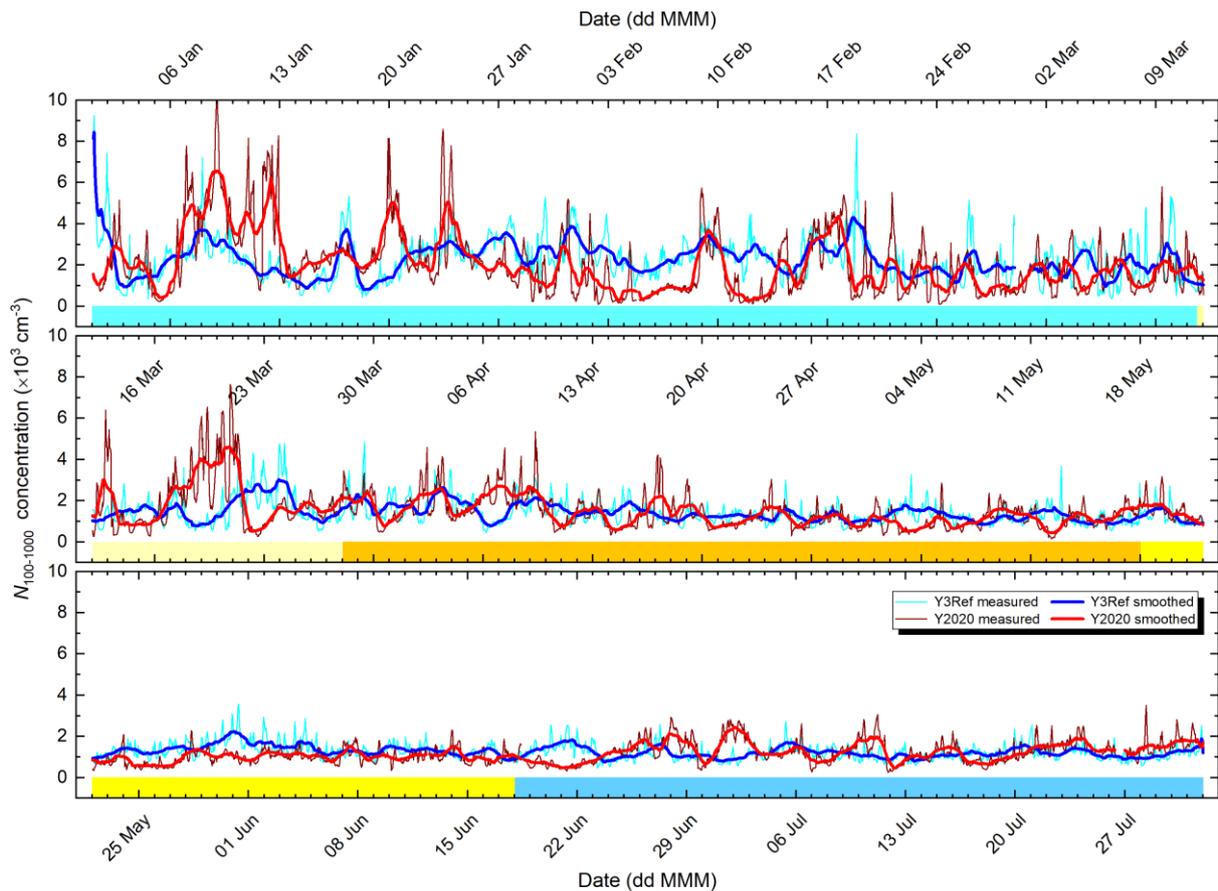
**Figure S7.** Time series of CO concentration in the average reference year of 2017–2019 (Y3Ref) and year 2020 together with their 24-h smoothed curves over the five consecutive phases of the first COVID-19 outbreak. The phases are marked by the following colour codes: Pre-emergency phase lighter blue, Pre-restriction phase lighter yellow, Restriction phase orange, Post-restriction phase darker yellow and Post-emergency phase darker blue. The tick labels of the abscissa indicate the Mondays in 2020.



**Figure S8.** Time series of SO<sub>2</sub> concentration in the average reference year of 2017–2019 (Y3Ref) and year 2020 together with their 24-h smoothed curves over the five consecutive phases of the first COVID-19 outbreak. The phases are marked by the following colour codes: Pre-emergency phase lighter blue, Pre-restriction phase lighter yellow, Restriction phase orange, Post-restriction phase darker yellow and Post-emergency phase darker blue. The tick labels of the abscissa indicate the Mondays in 2020.



**Figure S9.** Time series of PM<sub>10</sub> mass concentration in the average reference year of 2017–2019 (Y3Ref) and year 2020 together with their 24-h smoothed curves over the five consecutive phases of the first COVID-19 outbreak. The phases are marked by the following colour codes: Pre-emergency phase lighter blue, Pre-restriction phase lighter yellow, Restriction phase orange, Post-restriction phase darker yellow and Post-emergency phase darker blue. The daily health limit of 50  $\mu\text{g m}^{-3}$  is indicated by a horizontal line in darker green colour. The tick labels of the abscissa indicate the Mondays in 2020.



**Figure S10.** Time series of  $N_{100-1000}$  concentration in the average reference year of 2017–2019 (Y3Ref) and year 2020 together with their 24-h smoothed curves over the five consecutive phases of the first COVID-19 outbreak. The phases are marked by the following colour codes: Pre-emergency phase lighter blue, Pre-restriction phase lighter yellow, Restriction phase orange, Post-restriction phase darker yellow and Post-emergency phase darker blue. The tick labels of the abscissa indicate the Mondays in 2020.

## Reference

Maheras, P., Tolika, K., Tegoulas, I., Anagnostopoulou, Ch., Szpirosz, K., Károssy, Cs., Makra, L.: Comparison of an automated classification system with an empirical classification of circulation patterns over the Pannonian basin, Central Europe, Meteorol. Atmos. Phys., <https://doi.org/10.1007/s00703-018-0601-x>, 2018.