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Investigation of PM_{2.5} pollution during COVID-19 pandemic in Guangzhou, China

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ABSTRACT

The COVID-19 pandemic has raised awareness about various environmental issues, including PM_{2.5} pollution. Here, PM_{2.5} pollution during the COVID-19 lockdown was traced and analyzed to clarify the sources and factors influencing $PM_{2.5}$ in Guangzhou, with an emphasis on heavy pollution. The lockdown led to large reductions in industrial and traffic emissions, which significantly reduced PM_{2.5} concentrations in Guangzhou. Interestingly, the trend of PM_{2.5} concentrations was not consistent with traffic and industrial emissions, as minimum concentrations were observed in the fourth period (3/01-3/31, 22.45 μ g/m³) of the lockdown. However, the concentrations of other gaseous pollutants, e.g., SO₂, NO₂ and CO, were correlated with industrial and traffic emissions, and the lowest values were noticed in the second period (1/24-2/03) of the lockdown. Meteorological correlation analysis revealed that the decreased PM_{2.5} concentrations during COVID-19 can be mainly attributed to decreased industrial and traffic emissions rather than meteorological conditions. When meteorological factors were included in the PM_{2.5} composition and backward trajectory analyses, we found that long-distance transportation and secondary pollution offset the reduction of primary emissions in the second and third stages of the pandemic. Notably, industrial PM_{2.5} emissions from western, southern and southeastern Guangzhou play an important role in the formation of heavy pollution events. Our results not only verify the importance of controlling traffic and industrial emissions, but also provide targets for further improvements in PM_{2.5} pollution.

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Introduction

Guangzhou, with a permanent population of over 15.3 million in 2019 (Guangzhou Statistics Bureau, 2020), is a national central city and an international trade hub. Annual PM_{2.5} concentrations in Guangzhou decreased from 52.0 μ g/m³ in 2013 to 23.0 μ g/m³ in 2020, yet still exceed WHO Air Quality Guidelines (annual mean: 10 μ g/m³) (Guangzhou Municipal Ecological Environment Bureau, 2020). Moreover, the Guangdong-Hong Kong-Macao Greater Bay Area (GBA) has the worst air quality, including PM_{2.5} pollution, of the world's four major bay areas. On the premise of maintaining rapid economic growth, an accurate analysis of PM_{2.5} sources is critical for further improvements to the air quality in Guangzhou.

At the start of the COVID-19 pandemic, nationwide restrictive measures, such as stay-at-home recommendations, travel bans, cessation of public transportation, and the closing of shopping centers and entertainment venues, provided a unique opportunity to explore the dynamics and sources of PM_{2.5} contamination. This was done by comparing changes in PM_{2.5} concentrations and composition during different periods of the COVID-19 pandemic (Liu et al., 2020; Lv et al., 2020; Tanzer-Gruener et al., 2020). The degree of local PM_{2.5} contamination is determined by meteorological conditions, long-range transportation and atmospheric chemistry, as well as local emissions, with these factors exerting synergistic effects on PM_{2.5} pollution (An et al., 2019; Chen et al., 2020b; Wang et al., 2015; Zhao et al., 2013). The dramatic decrease in anthropogenic emissions, such as industrial and traffic emissions, during the COVID-19 lockdown greatly reduced the complexity of PM_{2.5} sources. This made it easier to accurately identify PM2.5 sources, as well as their relative contribution to pollution episodes. Nevertheless, unexpected heavy PM_{2.5} pollution, also termed "pandemic haze", was observed in several Chinese cities and regions, especially the Beijing-Tianjin-Hebei region (BTH), during different stages of the COVID-19 pandemic (Chang et al., 2020; Huang et al., 2021; Lv et al., 2020; Zhao et al., 2020b).

The pillar industries of GBA are mainly light industries such as electronics, electrical machinery and petrochemical industry; For this reason, the PM2.5 concentrations and composition in Guangzhou differ significantly from other regions in China, i.e., lower concentrations but a larger organic fraction (Chen et al., 2020a; Liao et al., 2021). Over the past year, numerous reports have analyzed PM2.5 contamination profiles, formation mechanisms and sources during the COVID-19 lockdown through various methods, e.g., satellite remote sensing, online monitoring, and mathematical models. These studies have focused on clarifying the inorganic components of PM_{2.5}, e.g., sulfate, nitrate, and ammonium (SNA), elemental carbon (EC), and crustal elements (CM) (Chang et al., 2020; Ghahremanloo et al., 2021; Li et al., 2020; Liu et al., 2020; Lv et al., 2020; Tanzer-Gruener et al., 2020). However, changes in the PM2.5 pollution levels in Guangzhou, especially those concerning the organic fraction, were rarely analyzed due to a lack of samples representing different phases of the COVID-19 pandemic.

In this study, main air pollutants ($PM_{2.5}$, CO, SO₂, NO_{2} , and O_{3}) and meteorological conditions were traced in Guangzhou

from 13 January to 30 April 2020, which were divided into five distinct periods according to the epidemic process. We quantitatively investigated the specific effects of both industrial and traffic emissions reduction due to COVID-19 lockdown and the variation of meteorological conditions on PM_{2.5} contamination in Guangzhou, a modern city based on the light industry in southern China. Furthermore, we tried to clarify the potential sources in Guangzhou based on the mass and composition of PM_{2.5} particles collected during the COVID-19 lockdown and backward trajectory analyses, with an emphasis on the heavy pollution cases. Our results can provide both targets and regulation strategies for further improvements in PM_{2.5} pollution.

1. Materials and methods

1.1. Sample and data collection

PM_{2.5} sampling campaigns were conducted in the Guangzhou Higher Education Mega Center (HEMC, 23.04°N, 113.37°E), which is a relatively independent area surrounded by the Pearl River. This sampling site was selected because of the following factors: (1) HEMC is located in the main city zone of Guangzhou and is thus representative of the local air quality; (2) heavy industrial and serious traffic emissions do not exist at the HEMC; therefore, the samples were assumed to reflect overall urban air conditions. Two medium-volume air samplers (Laoying Co. Ltd., Qingdao, China) with Quartz microfiber filters (Whatman, QMA, 90 mm diameter) were used to collect samples over 24 hr at a flow rate of 100 L/min. In this study, 24 PM_{2.5} samples were collected during the COVID-19 outbreak (8 samples/month) for quantitative analyses.

Data concerning local air pollutants ($PM_{2.5}$, SO_2 , NO_2 , CO and O_3) and meteorological conditions including temperature, relative humidity, rainfall, solar radiation, wind direction, wind speed and planetary boundary layer height (PBLH) were obtained from monitoring stations in HEMC, which was 1 km away from $PM_{2.5}$ sampling site, established by the Guangzhou Municipal Ecological Environment Agency and Guangzhou Meteorological Service.

The traffic flow is based on the statistics of 9 key toll stations in Guangzhou. Public traffic refers to the number of passengers on public transport. The statistics from the website of Guangzhou Municipal Transport Bureau (http://jtj.gz. gov.cn/jtzt/jtsj/jtysyb/index.html). The data of gross output value and added value of industrial enterprises above designated size in Guangzhou was obtained from the website of Guangzhou Bureau of Statistics. (http://112.94.72.17/portal/ queryInfo/macroReport/economicSituationIndex).

1.2. Time period setting

Quantifying changes in air quality during the COVID-19 lockdown requires precisely defined time periods that enable comparisons of spatiotemporal variations in $PM_{2.5}$ concentrations. The studied period was from January 13, 2020 to April 30, 2020. The COVID-19 epidemic in Guangdong province began on January 23 and the First-Level Public Health Emergency Response was initiated, which lasted until February 23. And then the emergency response to the novel coronavirus epidemic was

Table 1 – Division of the COVID-19 epidemic in Guangzhou into distinct time periods.			
Time period	COVID-19 epidemic	Corresponding time period in 2019	Information about the period
COVID-19 I	1/13-1/22	1/25-2/3	Before Chinese New Year Vacation and COVID-19 lockdown
COVID-19 II	1/23-2/03	2/04-2/10	Chinese New Year Vacation
COVID-19 III	2/04-2/29	2/11-2/28	Primary emergency response to the novel coronavirus epidemic
COVID-19 IV	3/01-3/31	3/01-3/31	Secondary emergency response to the novel coronavirus epidemic
COVID-19 V	4/01-4/30	4/01-4/30	After COVID-19 lockdown

lowered to the second level. By the end of March, the resumption rate of work and production of industrial enterprises above the scale in Guangdong Province was over 99% and the industrial production has basically returned to normal. Detailed information was summarized in Table S1. Considering the two factors of the local investigation in Guangzhou and Chinese New Year Vacation (CNY), we have divided the epidemic process into five distinct time periods (COVID-19 I to V, detailed in Table 1), and determined the corresponding time periods in 2019.

1.3. PM_{2.5} composition analysis

 $PM_{2.5}$ constituents were characterized according to a previously described methodology (Qi et al., 2020; Zhang et al., 2019). Water soluble inorganic ions, 16 Environmental Protection Agency (EPA) polycyclic aromatic hydrocarbons (PAHs), crustal elements (CM), organic carbon (OC) and elemental carbon (EC) were measured in the collected $PM_{2.5}$ samples during the COVID-19 epidemic. At first, the extraction of eight water-soluble inorganic ions (K⁺, Ca²⁺, Na⁺, Mg²⁺, Cl⁻, SO₄²⁻, NO₃⁻, NH₄⁺) was operated by the ultrasonic method. Quartz microfiber filters with $PM_{2.5}$ samples were fixed in a glass bottle with 15 mL ultrapure water and then took the ultrasound for 20 min with low temperature. After centrifuge, the supernatant was measured by ion chromatography (Aquion, Thermo Scientific, USA) to determine the concentrations of inorganic ions.

Sixteen priority PAHs, including naphthalene (Nap), acenaphthene (Ace), acenaphthylene (Acy), fluorene (Flu), phenanthrene (Phe), anthracene (Ant), fluoranthene (Flt), pyrene (Pyr), benz[a]anthracene (BaA), chrysene (Chr), benzo[b]fluoranthene (BbF), benzo[k]fluoranthene (BkF), Benzo[a]pyrene (BaP), indeno[1,2,3–cd]pyrene (IcdP), dibenz[a,h]anthracene (DahA), and benzo[ghi]perylene (BghiP) were extracted, concentrated, and analyzed. The analysis was operated on a TSQ 8000 Evo gas chromatograph mass spectrometer (Thermo Scientific, USA) equipped with a Thermo TRACETM1300 gas chromatograph, an electron ionization (EI) source, triple quadrupole analyzer and automatic injector (GC-EI-MS/MS). A Thermo TG-5MS capillary column (Thermo Scientific, USA) was applied to the separation.

The temperature program for the oven was as follows: the initial temperature of 80 °C was maintained for 1 min, and then increased to 180 °C at a rate of 5 °C/min (maintained for 2 min); the temperature continued increasing to 240 °C at a rate of 2.5 °C/min and was eventually increased to 300 °C at 3 °C/min and held for 1 min. The injections were splitless, and

the sample volume was 1 μ L. High-purity helium was used as the carrier gas at a constant flow rate of 1 mL/min. The temperature of the injector, ion source and transfer line were set at 280, 250 and 280 °C, respectively. Quantitative analysis was conducted on selected ion monitoring (SIM) mode. The chromatograph peaks of the samples were identified by mass spectra and retention time.

Inductively coupled plasma mass spectrometer (ICP-MS, 7700X; Agilent Technologies, Santa Clara, California, USA) was used to determine the elemental compositions of the collected samples on filters, including Na, Mg, Al, K, Ca, Ti, V, Cr, Ni, Mn, Fe, Ba, Cu, Zn, As and Pb. Carbonaceous aerosol components, OC and EC, were quantified using a Desert Research Institute (DRI) Model 2001 carbon analyzer (Atmoslytic Inc., Calabasas, CA, USA). The IMPROVE-A thermal/optical reflectance (TOR) protocol was used for the analyses (Chow et al., 2007; Wu et al., 2018).

1.4. Source appointment

The PAH diagnostic ratio represents a semi-quantitative way to identify sources of PAHs and the usefulness of PAH isomer ratios in source identification has been extensively proved (Dong et al., 2021; Le et al., 2020; Lu et al., 2017; Xu et al., 2020a). The diagnostic ratios of Ant/(Phe + Ant), Flt/(Flt + Pyr), IcdP/(IcdP + BghiP), and BaA/(BaA + Chr) were calculated to investigate the sources of PM2.5-bound PAHs during COVID-19 III to V (Table S2). The ratio of Ant/(Phe + Ant) <0.1 indicates the source of petroleum, while a ratio > 0.1 indicates a dominance of combustion(i.e. combustion of organic matter, anthropogenic industrial activities, or natural fire). Moreover, a ratio of Flt/(Flt + Pyr) < 0.4 stands for the petrogenic sources, 0.4–0.5 for petroleum combustion (especially liquid fossil fuel, vehicle and crude oil), while > 0.5 for combustion of biomass and coal. Ratios of IcdP/(IcdP + BghiP) < 0.2 and BaA/(BaA + Chr) < 0.20 indicate a petroleum source. Ratios of BaA/(BaA + Chr) of 0.20-0.35 and IcdP/(IcdP + BghiP) of 0.20-0.50 indicated origin of PAHs from petroleum combustion (liquid fossil fuel, vehicle, and crude oil combustion). If IcdP/(IcdP + BghiP) >0.50 and BaA/(BaA + Chr) > 0.50, the PAHs originated from coal, grass, and wood.

The air-mass backward trajectories were calculated and clustered to track the transport pathways of airflow arriving in HEMC (23.04°N, 113.37°E) using the Hybrid Single Particle Lagrangian Integrated Trajectory (HYSPLIT) model. The 72 hr backward trajectories started at 0:00 (UTC) for three distinct pandemic haze events (PEs) and were calculated at 500 m above ground level (AGL) using meteorological data



Fig. 1 – Changes in the concentration of AQI and five air pollutants during the COVID-19 lockdown and corresponding periods of 2019 in Guangzhou. * and \blacktriangle represent events with PM_{2.5} concentrations above 35 μ g/m³and below 12 μ g/m³, respectively.

available at the National Oceanic and Atmospheric Administration (NOAA) Global Data Assimilation System (GDAS).

1.5. Statistical analysis

Air pollutant concentrations (PM_{2.5}, SO₂, NO₂, CO and O₃) and meteorological conditions were expressed as a daily average value. The wind direction used in this study is the main wind direction of the day in question. The significance of differences between spatial and temporal changes in pollutant concentrations were determined using an unpaired Student's ttest. Pearson's correlation coefficients were calculated in SPSS software (Version 26.0, Chicago, IL, USA) to determine the statistical significance of relationships between PM_{2.5} pollution and meteorological parameters, with p < 0.05 set as the threshold for significance.

2. Results and discussion

2.1. Changes in air quality during the COVID-19 lockdown

Variations in the air quality of Guangzhou during different COVID-19 periods are shown in Fig. 1 and Table S3. SO₂, NO₂ and CO concentrations showed a V-shaped trend, i.e., an initial decrease followed by an increase, from period I to period V, with the lowest values occurring in period II (1/24-2/03, SO₂: 7.00 μ g/m³, NO₂: 15.64 μ g/m³, CO: 0.64 mg/m³), which is consistent with industrial and traffic emissions. Due to the close relationship between anthropogenic emissions and air quality, we also investigated how the COVID-19 lockdown affected local industry and traffic. The gross output value of industrial enterprises and number of cars on the road showed minimum values in February, i.e., 23.1% and 71% decreases, respectively, from the same period in February 2019 (Fig. 2, Ta-

bles S4, S5). However, O_3 concentrations increased by 9.15%, to 102.12%, across periods II, III, and V; this result may be explained by the reduced NO₂ that hinders the reaction between NO and O₃, thus increasing the atmospheric oxidizing capacity (Le et al., 2020; Xu et al., 2020a). Furthermore, we found that O_3 slightly decreased in stage IV in 2020. This result could be attributed to the short sunshine duration, lower solar radiation, and higher relative humidity (Table S6), which were unfavorable to the formation and accumulation of O₃ (Bu et al., 2021; Mousavinezhad et al., 2021; Yin et al., 2019).

During the fifth stage (COVID-V), industrial and agricultural production, along with transportation levels, in most of China, including Guangdong Province, had almost returned to normal, which was reflected in a significant increase in air pollutant concentrations when compared to the corresponding period in 2019 and previous COVID-19 phases.

Across periods I to IV, the average and median PM_{2.5} concentrations showed a downward trend, and the number of days with minimal PM_{2.5} concentrations ($\leq 12 \ \mu g/m^3$) increased. However, the degree to which PM_{2.5} concentrations dropped was still far from expectations. When compared with industrial emission trends, the reduction in PM_{2.5} concentrations showed a certain degree of lag. Notably, the unexpected increase in daily PM_{2.5} mainly appeared in the BTH region, yet severe haze pollution was observed over eastern China during the COVID-19 shutdown (Chang et al., 2020; Huang et al., 2021; Lv et al., 2020; Zhao et al., 2020b). This high frequency of haze events with high PM2.5 concentrations during the epidemic was mainly attributed to the higher secondary aerosol fraction of PM_{2.5} caused by the unbalanced reduction of gaseous pollutants (NO_X, O_3 and VOCs) along with medium-scale regional transportation (Ghahremanloo et al., 2021; Huang et al., 2021; Jia et al., 2020; Lv et al., 2020; Shen et al., 2021). The variation in PM_{2.5} concentrations over different regions during COVID-19 indicates that heavy industrial emissions are relevant to regional PM_{2.5} pollution.



Fig. 2 – The variations of industry and traffic during the COVID-19 lockdown in Guangzhou. (a), (b) The total output value and the growth rate of the three pillar industries in Guangzhou compared with the same period last year, (c) traffic congestion index and hourly road speed (Km/hr) in 2020 compared with the corresponding periods of 2019 in Guangzhou. Traffic congestion index and hourly road speed (Km/hr) in 2020.

The conversion of gaseous pollutants into $PM_{2.5}$ through photochemical reactions can be extended to precursors of organic gases, such as sulfate and nitrate, which can be transformed into secondary particulate matter through a series of chemical reactions (Huang et al., 2014; Sun et al., 2016). Pearson correlation coefficients were calculated for average daily PM_{2.5} and primary gas pollutant (CO, NO₂, SO₂, O₃) concentrations in Guangzhou across the studied COVID-19 phases. PM_{2.5} and NO₂ showed the highest Pearson correlation coefficients (r = 0.68), followed by PM_{2.5} and SO₂ (r = 0.63), O₃ (r =0.53) and CO (r = 0.41) (Fig. S1). The correlation coefficients between PM_{2.5} and gas pollutants during the COVID-19 shutdown were consistent with the values from previous years. However, the correlation coefficients between PM_{2.5} and both NO₂ and SO₂ during periods I to III significantly increased relative to period V and the same period in 2019, which indicates that the formation of secondary PM_{2.5} increased during the COVID-19 shutdown. In addition, the ratio of PM_{2.5}/CO (Table S7), an indicator of secondary pollutants to primary emissions, also increased during the COVID-19 shutdown, especially in period III. This further confirms the remarkable increase in the secondary aerosol fraction of PM_{2.5} in Guangzhou during the COVID-19 pandemic.

2.2. Influence of meteorological conditions on $PM_{2.5}$ concentrations during the COVID-19 lockdown

Meteorological conditions impart a significant impact on PM_{2.5} concentrations, mainly contributing to the diffusion, regional transportation, and secondary production of PM2.5 (Chen et al., 2020b; Xu et al., 2020b). We first calculated Pearson correlation coefficients for PM2.5 pollution and meteorological parameters to understand the mechanisms underlying the inconsistent reduction in anthropogenic emissions and PM_{2.5} concentrations in Guangzhou. As shown in Fig. 3a, the wind speed had the strongest effect on PM2.5 concentrations in Guangzhou during the COVID-19 epidemic (r = -0.56; p < 0.05), followed by daily total solar radiation, relative humidity, rainfall, PBLH and temperature. According to the rose diagrams and their relevant data of wind frequency- PM2.5 and PM_{2.5}-wind speed-wind direction (Fig. 3b, Table S8), the occurrence frequency of southeast wind was the highest and PM_{2.5} concentration was always maintained at a relatively high level (39.2–58.0 μ g/m³) when the southeast wind blows with the speed less than 2 m/sec, indicating that potential PM_{2.5} emission sources may be existed in the southeast of the monitoring point. Additionally, southern and western winds showed stronger correlations with PM2.5 pollution than winds from other directions (Fig. 3b, Table S8), which indicates that the observed PM_{2.5} pollution may also come from these directions. We then compared certain meteorological parameters during the five tested COVID-19 periods and the corresponding time points of 2019 to investigate the extent to which meteorological conditions can explain the observed variations in PM_{2.5} concentrations. Apparent differences in meteorological conditions were primarily explained by PBLH, relative humidity and temperature, both of which showed weak correlations with PM_{2.5} concentrations (Fig. 3a, Table S6). Furthermore, We compared the meteorological data over the past five years and found no abnormal weather conditions during the COVID-

19 epidemic, especially unfavorable meteorological conditions to $PM_{2.5}$ diffusion (Table S9). Thus, the decreased $PM_{2.5}$ concentrations during COVID-19 can be mainly attributed to decreased industrial and traffic emissions rather than meteorological conditions.

2.3. Composition analysis and source appointment of PM_{2.5} in Guangzhou

PAHs, which originate from various emission sources, are one of the most abundant organic components contributing to PM_{2.5} (Qi et al., 2020). Different PAH diagnostic ratios are commonly used to investigate possible sources of PAHs, and subsequently, PM_{2.5} (Dong et al., 2021; Gao and Ji, 2018; Yan et al., 2019). The Flt/(Flt + Pyr), BaA/(BaA + Chr) and Ant/(Ant + Phe) ratios were mostly higher than 0.50, 0.35 and 0.10, respectively, during the COVID-19 III period, revealing biomass and coal combustion, along with petroleum products, as the main source of $PM_{2.5}$ (Fig. 4a, b). This was expected, as activities vital to people's basic needs and the operation of the city, such as power generation and the production of certain essential materials, continued in lieu of social activities and transit (Zhao et al., 2020c). During the IV stage, the Flt/(Flt + Pyr) ratio varied was mostly > 0.50 (range: 0.50-0.64), while the Ant/(Ant + Phe) ratio was >0.10, which suggested biomass, coal, and petroleum combustion as a possible source of $PM_{2.5}$. During the same period, the BaA/(BaA + Chr) ratio was mostly > 0.35 (range: 0.31–0.37), while the IcdP/(IcdP + BghiP) ratio was <0.1. These ratios suggest a mixed source of PM_{2.5} (vehicle emissions, biomass and coal combustion, and petroleum sources) during the IV stage, which can be attributed to increased traffic and the reopening of industries, especially the petrochemical industry. During stageV, the BaA/(BaA + Chr) ratio was mostly < 0.35, whereas the Ant/(Ant + Phe) was > 0.10, revealing petroleum combustion, especially vehicle exhaust and energy production from liquid fossil fuel and crude oil, as a source of $PM_{2.5}$. Our results reveal that the change in PM_{2.5} composition was consistent with changes in traffic intensity. Since Guangzhou is characterized by light industry, PM_{2.5} pollution from industrial sources is less severe than that in the BTH region (Wang et al., 2020; Zhao et al., 2020a). In northern China, stagnant airflow and uninterrupted emissions from power plants and petrochemical facilities contributed to severe haze formation (Le et al., 2020). Conversely, transportation sources should also contribute more to total PM_{2.5} pollution in Guangzhou than what has been measured in other urban areas dominated by heavy industry. A similar study tracing the PM_{2.5} pollution during the COVID lockdown in Hangzhou, a southern city in China, also revealed that reductions in vehicular emissions were more responsible for the PM_{2.5} decline compared with stationary emissions (Liu et al., 2021). Therefore, traffic restrictions, which could be extensively studied during the COVID-19 lockdown, were effective at controlling PM_{2.5} pollution in Guangzhou.

Interestingly, the COVID-19 outbreak included three distinct pandemic haze events (PEs), with $PM_{2.5}$ concentration peaks occurring on 11 February (PE-1), 15 March (PE-2), and 8 April (PE-3). We determined the chemical composition of $PM_{2.5}$ during these three PEs, corresponding to COVID III, IV, and V, respectively (Fig. 5). The contribution of EC to $PM_{2.5}$ ini-



Fig. 3 – (a) The relationship between PM_{2.5} pollution and meteorological elements during the COVID-19 lockdown in Guangzhou, (b) The frequency distributions of wind directions (Left) and speeds (Right) with PM_{2.5} concentration (color demarcation) during the epidemic period of COVID-19. Wind speed: The wind speed represented by each grid increases progressively from the inside to the outside, increasing by 0.5 m/sec.

tially decreased and then increased from COVID-19 III to V. Since EC is formed by the inadequate combustion of carbonbased fuels, it can usually be used to identify primary emission sources (Wu et al., 2018). Therefore, the initial decrease in the share of EC in PM2.5 indicates a decrease in primary sources of PM_{2.5} during the lockdown periods. In addition, secondary pollutants (OC, SO₄²⁻, NO₃⁻ and NH₄⁺) were measured at especially high concentrations in comparison to the COVID-V period, suggesting that PM_{2.5} pollution during the COVID-19 lockdown represented the enhanced formation of secondary aerosols with increased atmospheric oxidizing capacity. This is in agreement with previous research, as numerous studies have reported that the rapid formation of secondary inorganic aerosols was the main factor contributing to air pollution during the COVID-19 outbreak (Chang et al., 2020; Huang et al., 2021; Nichol et al., 2020).

The formation of haze is also strongly linked to the regional transportation of $PM_{2.5}$. To evaluate how regional transportation influences $PM_{2.5}$ pollution, a three-day back trajectory cluster analysis at 500 m was performed for each PEs, with the average $PM_{2.5}$ concentrations in the trajectory clusters

shown in Table S10. The results revealed several transportation pathways for the studied PEs. The average PM_{2.5} concentrations in two airflow trajectories entering Guangzhou from the northeast and southeast (Cluster 1 represents local emissions, while Cluster 2 represents marine transportation) are relatively high (36.5 μ g/m³, 36.3 μ g/m³), indicating that local sources significantly impacted PM_{2.5} pollution in Guangzhou during COVID-19 period III (Fig. 4c). During the next period, air masses (Cluster 1 and Cluster 3) from the northwest, which passed over heavily polluted regions in northern and central China (e.g., Shanxi, Henan, and Anhui provinces), showed the highest PM_{2.5} levels (37.3 μ g/m³) (Fig. 4d). This indicates that PM_{2.5} pollution in Guangzhou was also affected by longrange transportation during the later stages of the COVID-19 lockdown. During phase V, air masses from the northeast of Guangzhou (Cluster 1), which represented the shortest inland trajectory, accounted for 50 % of total PM2.5 pollution and also showed the highest average concentration of PM2.5 (36.1 μ g/m³). Cluster 2 accounted for 25% of PM_{2.5} pollution, with an average concentration of 31 μ g/m³ (Fig. 4e). Nevertheless, long inland trajectories from northeast China (Clus-



Fig. 4 – Cross plot for the PAH ratios of (a) Ant/(Ant + Phe) vs. Flt/(Flt + Pyr) and (b) BaA/(BaA + Chr) vs IcdP/(IcdP + BghiP), (c-e) Cluster analysis of air mass back trajectories during the COVID-19 lockdown.



Fig. 5 – $PM_{2.5}$ concentration and composition of three pandemic haze events. OM (organic matter) = 1.6 *OC, EC: elemental carbon, SNA: sulfate, nitrate, and ammonium, CM (crustal elements) = 1.94 Ti + 2.2 Al + 1.63 Ca + 2.42 Fe + 2.49 Si, others = $PM_{2.5}$ mass–OM–EC–SNA–CM.

ters 3 and 4) account for a relatively low percentage (both 13%) and carry low average concentrations of PM_{2.5} ($21 \,\mu g/m^3$, 17.7 $\mu g/m^3$). Hence, PM_{2.5} pollution in Guangzhou mainly originated from local sources during the COVID-19 V period. The backward trajectory analysis showed that PM_{2.5} pollution in Guangzhou mainly originated from local sources during period III and V, and resulted from long-distance transportation during stages IV. This corroborates what has been reported in other recent studies, i.e., local emissions and regional pol-

lutant transportation most likely caused the pandemic haze events observed during the COVID-19 pandemic (Li et al., 2020; Shen et al., 2021; Zhao et al., 2020b).

We also studied the prevailing wind direction(s) to identify emission sources. The results revealed that: (1) southeasterly winds prevailed during the study period, while the westerly, southerly and southeasterly winds always showed high $PM_{2.5}$ concentrations; (2) most of the polluting industries (petrochemical industry and power plants) are located to the west, southerly and southeast of Guangzhou and its surrounding cities (Foshan, Zhaoqing, Dongguan and Huizhou) (Fig. S2). Based on these observations, it can be concluded that westerly, southerly and southeasterly winds will bring pollutants to Guangzhou and aggravate local $PM_{2.5}$ pollution. Therefore, controlling the emissions by these industries is of paramount importance to improving the air quality in Guangzhou.

3. Conclusions

Measurements taken during the COVID-19 lockdown show that PM_{2.5} pollution in Guangzhou improved significantly as traffic and industrial emissions fell, with PM2.5 levels below 6 μ g/m³ observed across five days of COVID-IV. These results indicate that controlling traffic and industrial emissions may be decisive for PM_{2.5} pollution in Guangzhou; this is not the case for the BTH region and several cities in the Yangtze River Delta. Spatiotemporal analyses based on wind direction and $\text{PM}_{2.5}$ distribution indicate that decision-makers must also pay attention to PM_{2.5} emissions originating from industrial areas to the south, southeast and west of Guangzhou, which can arrive by long-range transportation. The results also showed that a sharp decrease in primary PM2.5 emissions can lead to an increase in the production of secondary particles, which - along with the long-distance transportation of particles - can offset the initial decrease in primary PM_{2.5} emissions. Additionally, O₃ was the only atmospheric pollutant that demonstrated increased levels during COVID-19; hence, O₃ pollution will pose a challenge for Guangzhou and other regions in China. The present results confirm the effectiveness of previous PM_{2.5} control measures in Guangdong Province, e.g., reducing coal combustion, construction site dust diffusion, and emissions linked to manufacturing, as well as installing monitoring equipment; on the other hand, the performed analyses also identified southeastern and western zones of Guangzhou as significant sources of PM_{2.5}. Moreover, the research highlights how ozone pollution can increase when PM_{2.5} levels fall.

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Appendix A Supplementary data

Supplementary material associated with this article can be found, in the online version, at doi:10.1016/j.jes.2021.07.009.

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