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Air quality characteristics during 2016–2020 in Wuhan, China

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Implementation of a clean air policy in China has high national importance. Here, we analyzed tempo-spatial characteristics of the concentrations of PM_{2.5} (PM_{2.5-C}), PM₁₀ (PM_{10-C}), SO₂ (SO_{2-C}), NO₂ (NO_{2-C}), CO (CO-C), and maximum 8-h average O₃ (O_{3-8h-C}), monitored at 22 stations throughout the mega-city of Wuhan from January 2016 to December 2020, and their correlations with the meteorological and socio-economic factors. PM_{2.5-C}, PM_{10-C}, SO_{2-C}, NO_{2-C}, and CO-C showed similar monthly and seasonal trends, with minimum value in summer and maximum value in winter. However, O_{3-8h-C} showed an opposite monthly and seasonal change pattern. In 2020, compared to the other years, the annual average PM_{2.5-C}, PM_{10-C}, SO_{2-C}, NO_{2-C}, and CO-C were lower. PM_{2.5-C} and PM_{10-C} were higher in urban and industrial sites and lower in the control site. The SO_{2-C} was higher in industrial sites. The NO_{2-C} was lower, and O_{3-8h-C} was higher in suburban sites, while CO showed no spatial differences in their concentrations. PM_{2.5-C}, PM_{10-C}, SO_{2-C}, NO_{2-C}, and CO-C had positive correlations with each other, while O_{3-8h-C} showed more complex correlations with the other pollutants. PM_{2.5-C}, PM_{10-C}, SO_{2-C}, and CO-C presented a significantly negative association with temperature and precipitation, while O₃ was significantly positively associated with temperature and negatively associated with relative air humidity. There was no significant correlation between air pollutants and wind speed. Gross domestic product, population, number of automobiles, and energy consumption play an important role in the dynamics of air quality concentrations. These all provided important information for the decision and policy-makers to effectively control the air pollution in Wuhan.

In past decades, the deterioration of air quality caused by increased human activity and manufacturing has attracted wide concern worldwide^{1,2}. This not only reduces the visibility of the air atmosphere³ but also significantly harms human health⁴ and endangers the sustainable development of society and the economy⁵. High concentrations of PM_{2.5} and PM₁₀ would reduce atmospheric visibility and increase the occurrence of traffic accidents, while excessive exposure to polluted air could cause many kinds of cardiovascular and chronic respiratory diseases (e.g. asthma)⁶ or even lead to premature death and cancer². High concentrations of SO₂ and NO_x could cause acid rain and bring serious adverse ecosystem effects, such as the corrosion of buildings, soil acidification, and damage to crops and the aquatic environment^{7,8}. Therefore, it is urgent to reduce air pollutants and improve air quality.

Urban air quality is influenced by various factors, including socio-economic factors (e.g., the level of economic development, urban population, car ownership, fuel emission, and usage of fossil resources, etc.) and meteorological factors (e.g., temperature, relative air humidity, wind speed, and precipitation, etc.)^{9,10}. Socio-economic factors are the main pollution sources affecting urban air quality^{11,12}, while meteorological conditions also influence air quality when the main pollution sources are relatively stable^{13–16}. For example, urban emissions from human activity and manufacturing could cause environmental problems, such as photochemical smog, ozone layer depletion, acid rain, toxic chemical pollution, and global climate warming^{17,18}. Monthly or seasonal air quality variations are caused by pollutant-intensive emission sources and meteorological conditions^{13,19,20}. Therefore, strict and effective regulations and measures for air pollution prevention have begun to be implemented worldwide^{21,22}.

China has suffered serious environmental degradation and air pollution, accompanying rapid economic growth and urbanization in recent decades. Population growth, energy consumption, motor vehicle increment,

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and industrial dust emission have become China's main causes of air pollution²³. Many studies have reported that air pollution strongly impacts people's health and life^{24,25}. In order to mitigate air pollution, many pollution reduction technologies and policies have been implemented to improve air quality. The control measures (e.g., renewable energy utilization, traffic control, and flue gas desulfurization and denitration) adopted by China's central and local governments have achieved remarkable results. However, air pollution is still very serious^{26,27}, and has gradually become a hot topic of high concern in China²⁸. Scientific evaluation of the temporal change and spatial differences in air quality characteristics can help the government to examine the air pollution status and dynamics, while systematic analysis of the factors affecting air quality could provide evidence for policymakers to formulate effective measures¹⁰, and to optimize the urban expansion and development pattern, and land use/land cover characteristics to improve air quality at city levels^{29–32}.

The temporal characteristics of air quality and their driving factors have been analyzed across multiple spatial scales in China, mainly in megacities (e.g., Beijing, Shanghai, Guangzhou, and Shenzhen, etc.)^{23,33–35}. However, the spatial heterogeneity in air pollutants has seldom been evaluated at city scales due to limited observation sites. As one of the important and core cities in China's Yangtze River Economic Belt, Wuhan's rapid industrialization and urbanization have achieved short-term gains at the expense of the environment. Assessing and exploring air quality in Wuhan will guide Hubei Province and even the Yangtze River Economic Belt to a certain extent. However, few studies have comprehensively examined the temporal characteristics of air quality in Wuhan city, especially the spatial variations characteristics.

In this study, the official data on daily Individual Air Quality Index (IAQI) of PM_{2.5}, PM₁₀, SO₂, NO₂, CO, and maximum 8-h average O₃ (O_{3_8h}), monitored at 22 stations throughout the mega-city of Wuhan from January 2016 to December 2020, were used to examine spatio-temporal characteristics of air pollution in Wuhan city. The main objectives of this study are: (1) evaluate the variations in the concentrations of PM_{2.5} (PM_{2.5_C}), PM₁₀ (PM_{10_C}), SO₂ (SO_{2_C}), NO₂ (NO_{2_C}), CO (CO_C), and O_{3_8h} (O_{3_8h_C}) at monthly, seasonal, and yearly scales in the air in Wuhan city during 2016–2020; (2) examine the spatial differences in PM_{2.5_C}, PM_{10_C}, SO_{2_C}, NO_{2_C}, CO_C, and O_{3_8h_C} at seasonal and yearly scales across different sites; and (3) analyze the influence of the main meteorological factors and socio-economic indicators on air pollutants in Wuhan city. The results should provide city-scale evidence on spatio-temporal characteristics of air pollutants and a scientific basis for taking effective measures and new policy proposals to improve air quality in Wuhan city.

Materials and methods

Study area. Wuhan has a subtropical monsoon, humid climate with four distinct seasons. The annual average precipitation is 1205 mm, and the annual average temperature is 15.8–17.5°C³⁶. As one of central China's most important cities and mega-cities, Wuhan has a land area of 8569.15 km². It is an important hub in China due to its many waterways, convenient transportation, and advantageous position. With the rapid rise in its scale, the central strategic fulcrum of Wuhan city has become increasingly significant. Rapid urbanization, high population growth, a large number of vehicles, and high energy consumption have all led to ecological degradation in recent decades in Wuhan³⁷.

Sampling site and air quality datasets. Data on the daily IAQI of PM_{2.5}, PM₁₀, SO₂, NO₂, CO, and O_{3_8h} from January 2016 to December 2020 at 22 monitoring stations were derived from the official website of the Wuhan ecological environment bureau, which is open-access (<http://hbj.wuhan.gov.cn/>). PM_{2.5_C}, PM_{10_C}, SO_{2_C}, NO_{2_C}, CO_C, and O_{3_8h_C} were calculated from the IAQI. The locations of 22 monitoring stations of the air quality in Wuhan were showed in Fig. 1. The figure is generated using ArcGIS 10.4 (ArcGIS for windows, version 10.4, Environmental Systems Research Institute, Inc., USA). The DEM data was downloaded from Geo-spatial Data Cloud (<https://www.gscloud.cn/sources/>).

The 22 stations include one control site (Chenhu Qihao), one industrial site (Huagong district), eleven urban central sites (Hankou Huaqiao, Hankou Jiangtan, Hanyang Yuehu, Jiangnan Honglingjin, South Jiangnan, Qiaokou Gutian, Wuchang Ziyang, Donghu Liyuan, No.182 Minzu Avenue, Hongshan Dida, Qingshan Ganghua), and nine suburban sites (Zhuankou district, Caidian district, Wujiashan, Dongxihu district, Huangpi district, Xinzhou district, Huashan Ecological Art Museum, Hannan district, and Jiangxia district). The control site was located in the national wetland reserve, representing the background level of air quality in Wuhan.

Meteorological and socio-economic datasets. Daily meteorological data of the Wuhan meteorological observation station (30°37' N, 114°08' E) were downloaded from the China Meteorological Data Sharing Service System (<http://data.cma.cn/>). The meteorological factors data mainly include air temperature (*T*, °C), relative air humidity (*RH*, %), precipitation (*Prec*, mm), and wind speed (*w*, m/s) from 2016 to 2020. The data on socio-economic factors were obtained from the Wuhan Statistical Yearbook (<http://tj.wuhan.gov.cn/>). The socio-economic factors data mainly include the yearly gross domestic product (*GDP*, 100 million yuan), per capita regional GDP (*PGDP*, yuan/population), permanent resident population (*PRP*, ten thousand population), population density (*PD*, population/km²), road area (*RA*, × 10⁴ m²), number of civilian vehicles (*CV*), per capita area of parks and green spaces (*PPGA*, m²), the green coverage rate of built-up areas (*GRB*, %), total energy combustion (*EC*, tons of standard coal/km²), coal combustion (*CAC*, tons of standard coal/km²), coke combustion (*CKC*, tons of standard coal/km²), crude oil combustion (*COC*, tons of standard coal/km²), fuel oil combustion (*FOC*, tons of standard coal/km²), and electric power combustion (*EPC*, kWh) from 2009 to 2020. The data from 2020 were excluded from the correlation analysis results due to COVID-19 in 2020.

Statistical analysis. Correlation analysis was used to analyze the relevance of the six pollutants and the influence of the main meteorological factors and socio-economic indicators on the six pollutants. Monthly and

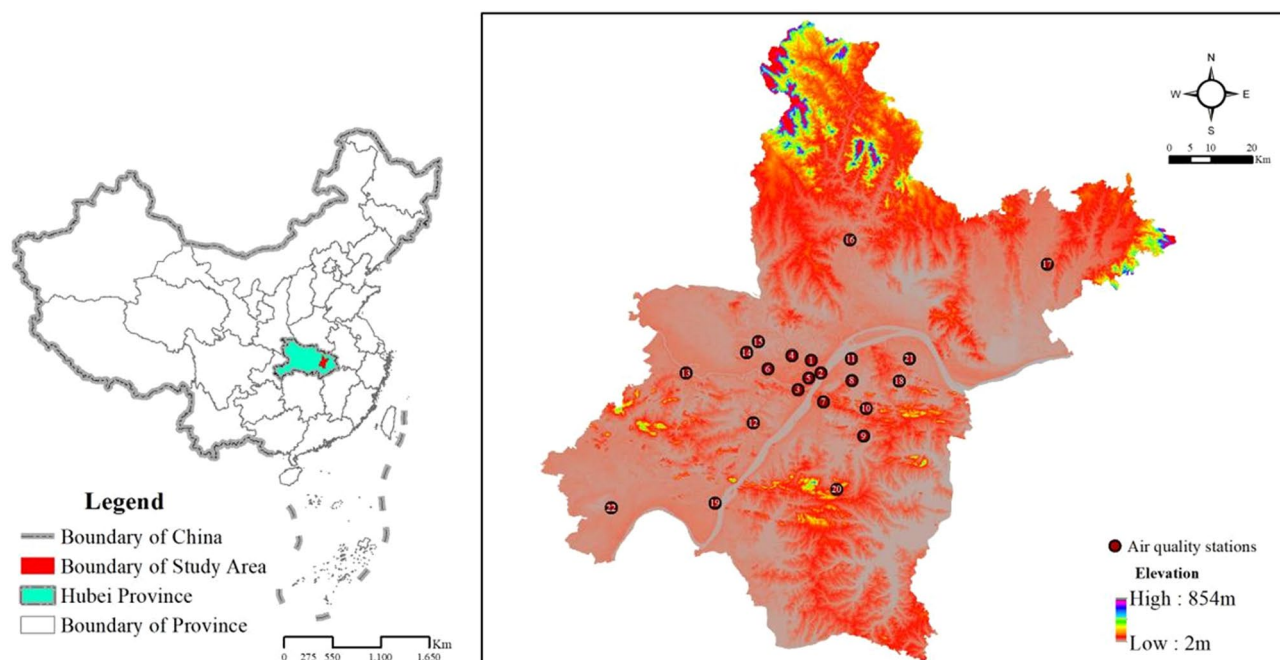


Figure 1. Locations of the air quality monitoring stations in Wuhan city. Notes: 1, Hankou Huaqiao; 2, Hankou Jiangtan; 3, Hanyang Yuehu; 4, Jiangnan Honglingjin; 5, South Jiangnan; 6, Qiaokou Gutian; 7, Wuchang Ziyang; 8, Donghu Liyuan; 9, No. 182 Minzu Avenue; 10, Hongshan Dida; 11, Qingshan Ganghua; 12, Zhuankou district; 13, Caidian district; 14, Wujiashan; 15, Dongxihu district; 16, Huangpi district; 17, Xinzhou district; 18, Huashan Ecological Art Museum; 19, Hannan district; 20, Jiangxia district; 21, Huangong district; 22, Chenhu Qihao.

yearly data were mainly used to analyze the correlation between air quality, meteorological, and socio-economic factors. Before correlation analysis, independence, and normality tests were conducted on the data. Pearson correlation coefficients were calculated for the relationships between air pollutants and the meteorological and socio-economic data. The correlation between factors was regarded as statistically significant or highly significant when the *P* value was less than 0.05 or 0.01, respectively. Detailed methods can be referred to Chen et al.²³. Data analysis was performed through SPSS 16.0 (SPSS for Windows, version 16.0, Chicago, Illinois, USA), and the spatial and temporal variations were plotted using SigmaPlot 10.0 (SigmaPlot for windows, version 10.0, San Jose, California, USA).

Results

Temporal variations in air pollutants. $PM_{2.5_C}$, PM_{10_C} , SO_{2_C} , NO_{2_C} , and CO_C showed similar monthly and seasonal trends. In contrast, $O_3_8h_C$ showed opposite monthly and seasonal trends (Fig. 2). At the monthly scale, $PM_{2.5_C}$, PM_{10_C} , SO_{2_C} , NO_{2_C} , and CO_C all showed a single-valley change pattern, with the minimum value in July and the maximum value in December or January. In comparison, $O_3_8h_C$ showed a double-peak change pattern, with the peak value in June and September (Fig. 2a). At the seasonal scale, the maximum values of $PM_{2.5_C}$, PM_{10_C} , SO_{2_C} , NO_{2_C} , and CO_C were in winter (December, January, and February) and the minimum values were in summer (June, July, and August), while the maximum value of $O_3_8h_C$ was in summer and the minimum value was in winter (Fig. 2b). At the yearly scale, the maximum values of the annual average $PM_{2.5_C}$, PM_{10_C} , SO_{2_C} , and CO_C were in 2016; the maximum value of the annual average NO_{2_C} was in 2017; while the maximum value of the annual average $O_3_8h_C$ was in 2019 (Fig. 2c). Annual average $PM_{2.5_C}$, PM_{10_C} , SO_{2_C} , NO_{2_C} , and CO_C were lower in 2020 compared to all the other years, which was mainly attributed to the strict lockdown in Wuhan in early 2020.

Spatial variations in air pollutants. The seasonal trends in $PM_{2.5_C}$, PM_{10_C} , SO_{2_C} , NO_{2_C} , CO_C , and $O_3_8h_C$ from 2016 to 2020 were similar at all 22 sites (Fig. 3), while the yearly trends in $PM_{2.5_C}$, PM_{10_C} , SO_{2_C} , NO_{2_C} , CO_C , and $O_3_8h_C$ from 2016 to 2020 showed small differences between the sites (Fig. 4). Overall, all the air pollutants showed distinct seasonal patterns at all sites (Fig. 3). $PM_{2.5_C}$, PM_{10_C} , SO_{2_C} , NO_{2_C} , and CO_C exhibited the lowest and highest concentrations in summer and winter, respectively, at all observation sites. In contrast to $PM_{2.5_C}$, PM_{10_C} , SO_{2_C} , NO_{2_C} , and CO_C , $O_3_8h_C$ exhibited the lowest concentrations in winter at all observation sites. During the observation year, $PM_{2.5_C}$ and PM_{10_C} decreased year by year at each station. The $PM_{2.5_C}$ was highest in Qingshan Ganghua. The PM_{10_C} was higher in Hankou Huaqiao, Jiangnan Honglingjin, Qiaokou Gutian, Qingshan Ganghua, Wujiashan, Dongxihu district, and Huangong district than in other stations. The SO_{2_C} and NO_{2_C} at several stations were highest in 2017 and lowest in 2020, while $O_3_8h_C$ was the highest in 2019 at some stations (Fig. 4). The NO_{2_C} was lower while

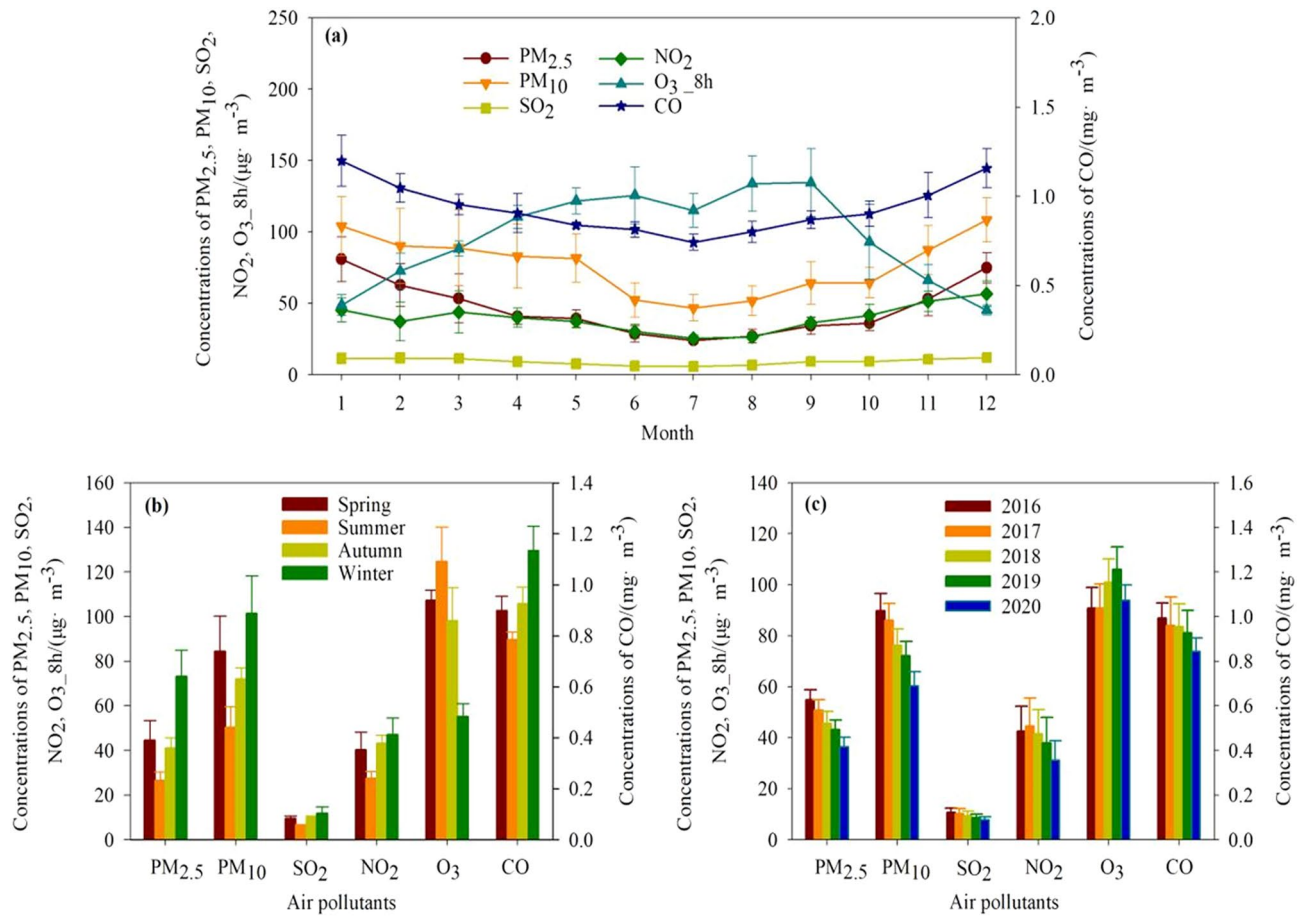


Figure 2. Temporal variations in PM_{2.5_C}, PM_{10_C}, SO_{2_C}, NO_{2_C}, CO_C, and O_{3_8h_C} at monthly (a), seasonal (b), and yearly (c) scales during 2016–2020.

O_{3_8h_C} and SO_{2_C} were higher in suburban sites than that in urban central sites. The PM_{10_C}, SO_{2_C}, NO_{2_C}, and CO_C were higher in suburban and urban central sites than in the control site (Fig. 4).

The correlation between air pollutants and other factors. Correlations between the six air pollutants. Significant correlations were found between the concentrations of the six air pollutants ($p < 0.01$), except for the relevance between O_{3_8h_C} and PM_{10_C} ($p > 0.05$) (Table 1). PM_{2.5_C}, PM_{10_C}, SO_{2_C}, NO_{2_C}, and CO_C showed significantly positive correlations with each other ($p < 0.01$), while O_{3_8h_C} presented significantly negative correlations with PM_{2.5_C}, SO_{2_C}, NO_{2_C}, and CO_C ($p < 0.01$).

Correlations between air pollutants and influencing factors. The correlation between Temperature (T) and O_{3_8h_C} was significantly positive ($p < 0.01$), while the correlations between the other pollutants and T were significantly negative ($p < 0.01$) (Table 2). Relative air humidity (RH) was negatively correlated with SO_{2_C} and O_{3_8h_C} ($p < 0.05$), while the correlations between the other pollutants and RH were insignificant ($p > 0.05$) (Table 2). Precipitation ($Prec$) was negatively correlated with all pollutants ($p < 0.01$) except O_{3_8h_C} (Table 2). There was no significant correlation between air pollutants and wind speed (w) ($p > 0.05$) (Table 2).

PM_{2.5_C}, PM_{10_C}, SO_{2_C}, NO_{2_C}, and CO_C presented a significantly negative relationship with GDP, PGDP, PRP, RA, CV, and GRB ($p < 0.05$ or $p < 0.01$). Significantly negative relationships were found between PM_{2.5_C}, SO_{2_C}, NO_{2_C}, CO_C, and PD ($p < 0.05$ or $p < 0.01$). PM_{2.5_C} displayed a significantly positive relationship with EC, CAC, CKC, and FOC. PM_{10_C} displayed a significantly positive relationship with FOC. SO_{2_C} displayed a significantly positive relationship with EC, CAC, CKC, COC, FOC, and EPC and NO_{2_C} displayed a significantly positive relationship with CAC, CKC, and FOC. There was no correlation between CO_C and EC, CAC, CKC, COC, FOC, EPC, and between O_{3_8h_C} and all the socio-economic indicators (Table 3).

Discussion

Temporal variations in air pollutants. The monthly and seasonal trends in PM_{2.5_C}, PM_{10_C}, SO_{2_C}, NO_{2_C}, CO_C and O_{3_8h_C} can reflect the formation processes of the air pollutants. According to the results, PM_{2.5_C}, PM_{10_C}, SO_{2_C}, NO_{2_C}, and CO_C all showed a single-valley change pattern, with the maximum value in winter and the lowest value in summer. The lower PM_{2.5_C}, PM_{10_C}, SO_{2_C}, NO_{2_C}, and CO_C in summer are mainly attributable to better wet removal and atmospheric boundary layer diffusion conditions

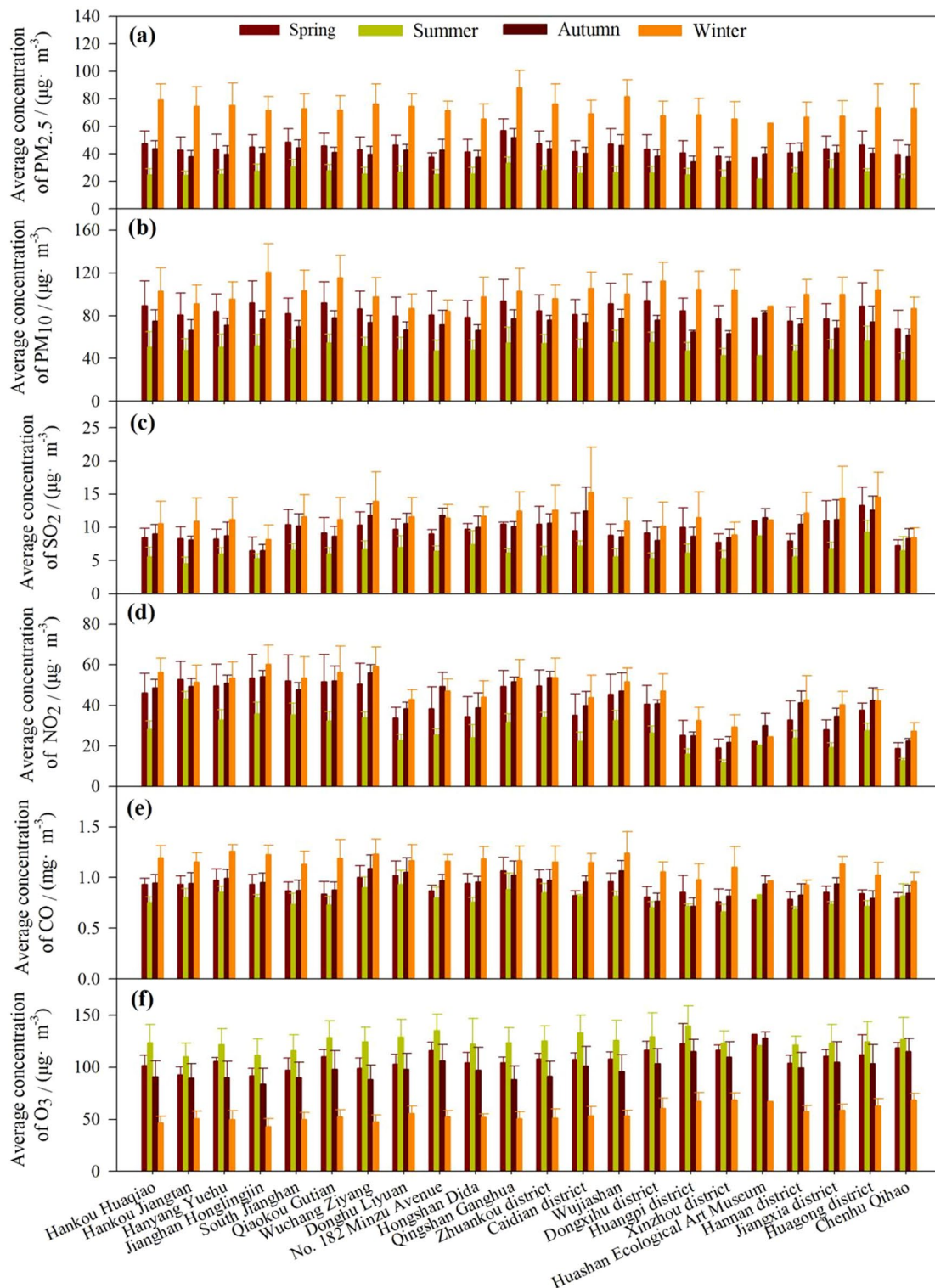


Figure 3. Spatial differences in PM_{2.5-C}, PM_{10-C}, SO_{2-C}, NO_{2-C}, CO_{-C}, and O_{3-8h-C} at seasonal scale during 2016–2020 across different sites.

(Fig. 5), and fewer combustion sources. In comparison, emissions from heat sources were important contributors to the higher PM_{2.5-C}, PM_{10-C}, SO_{2-C}, NO_{2-C}, and CO_{-C} in winter^{38,39} because of the lowest temperature (Fig. 5). O_{3-8h-C} showed a double-peak change pattern that mostly peaked in summer due to higher solar radiation and temperatures in summer time (Fig. 5) could contribute to the photochemistry activity and thus the formation of O₃^{40,41}.

The PM_{2.5-C}, PM_{10-C}, SO_{2-C}, NO_{2-C}, and CO_{-C} decreased year by year and were lowest in 2020, mainly attributed to the effectiveness of the environmental protection and pollution control strategies²³, and the strict

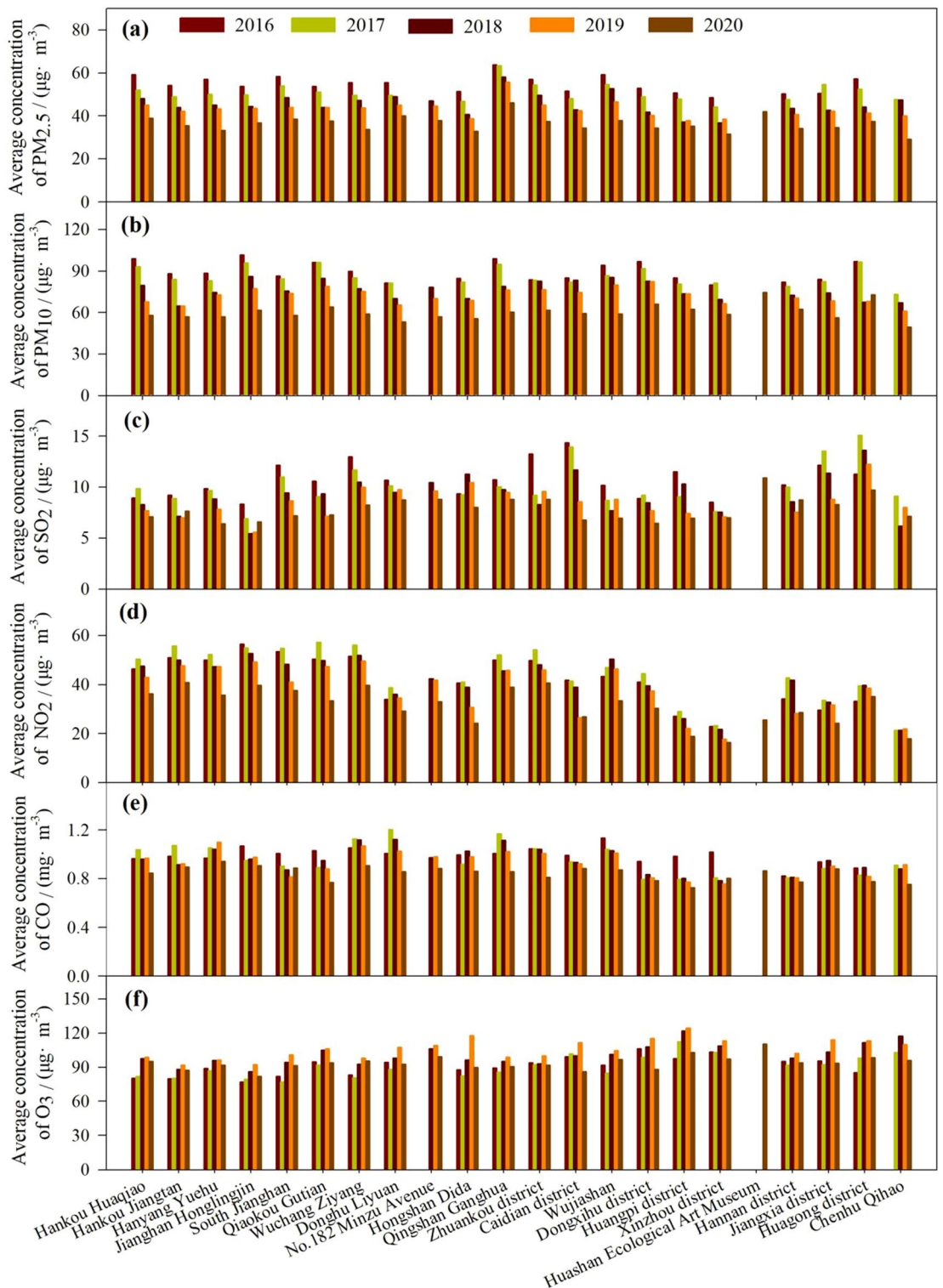


Figure 4. Spatial differences in PM_{2.5}-C, PM₁₀-C, SO₂-C, NO₂-C, CO-C, and O₃-8h-C at the yearly scale during 2016–2020 across different sites.

lockdown in Wuhan in early 2020. The phenomenon of pollutant reduction during the COVID-19 lockdown in this study is consistent with that reported in previous studies⁴². The decrease in SO₂-C in Wuhan city indicated an effective upgrading of key industrial sectors (electric power and steel, etc.), especially the ultra-low emission transformation of electric power, the elimination of small and medium-sized coal-fired boilers, the transformation of heating from coal to gas and electricity. This was consistent with the study by Li et al.²⁶, which showed that the total SO₂ emission in China in 2016 was 75% less than that in 2007. The decrease in CO-C and NO₂-C was mainly attributed to the effective regulation of traffic-related and coal combustion emissions. These all

	PM ₁₀ _C	SO ₂ _C	NO ₂ _C	CO_C	O ₃ _8h_C
PM _{2.5} _C	0.766**	0.478**	0.517**	0.635**	-0.219**
PM ₁₀ _C		0.492**	0.562**	0.458**	-0.007
SO ₂ _C			0.522**	0.383**	-0.016**
NO ₂ _C				0.499**	-0.135**
CO_C					-0.221**

Table 1. Correlations between the daily concentrations of the six pollutants during 2016–2019 (** $p < 0.01$; * $p < 0.05$).

	PM _{2.5} _C	PM ₁₀ _C	SO ₂ _C	NO ₂ _C	CO_C	O ₃ _8h_C
<i>T</i>	-0.901**	-0.749**	-0.717**	-0.768**	-0.889**	0.869**
<i>RH</i>	-0.067	-0.278	-0.357*	-0.097	0.116	-0.377**
<i>Prec</i>	-0.385**	-0.372**	-0.485**	-0.378**	-0.396**	0.164
<i>w</i>	0.092	0.099	-0.011	-0.198	-0.011	-0.122

Table 2. Correlations between the monthly concentrations of the six pollutants and meteorological factors during 2016–2019 (*T* temperature; *RH* relative air humidity; *Prec* precipitation; *w* wind speed; ** $p < 0.01$; * $p < 0.05$).

	PM _{2.5} _C	PM ₁₀ _C	SO ₂ _C	NO ₂ _C	CO_C	O ₃ _8h_C
<i>GDP</i>	-0.921**	-0.739**	-0.951**	-0.802**	-0.795*	0.707
<i>PGDP</i>	-0.918**	-0.725*	-0.953**	-0.794**	-0.788*	0.704
<i>PRP</i>	-0.978**	-0.663*	-0.942**	-0.720*	-0.779*	0.637
<i>PD</i>	-0.963**	-0.578	-0.835**	-0.686*	-0.789*	0.656
<i>RA</i>	-0.920**	-0.815**	-0.927**	-0.841**	-0.841*	0.731
<i>CV</i>	-0.950**	-0.789**	-0.940**	-0.848**	-0.796*	0.691
<i>PPGA</i>	0.512	0.099	-0.598	-0.153	0.755	-0.547
<i>GRB</i>	-0.851*	-0.470	-0.941**	-0.685*	-0.676	0.600
<i>EC</i>	0.917**	0.476	0.940**	0.597	0.406	-0.159
<i>CAC</i>	0.842*	0.536	0.960**	0.686*	0.449	-0.247
<i>CKC</i>	0.909**	0.480	0.947**	0.626*	0.518	-0.418
<i>COC</i>	-0.072	0.142	0.606*	0.313	0.050	0.084
<i>FOC</i>	0.803*	0.709*	0.834**	0.757**	0.516	-0.176
<i>EPC</i>	0.580	0.378	0.873**	0.548	0.125	0.068

Table 3. Correlations between the yearly concentrations of the six pollutants and socio-economic indicators during 2009–2019 (*GDP* gross domestic product; *PGDP* per capita regional *GDP*; *PRP* permanent resident population; *PD* population density; *RA* road area; *CV* number of civilian vehicles; *PPGA* per capita area of parks and green spaces; *GRB* green coverage rate of built-up areas; *EC* total energy combustion; *CAC* coal combustion; *CKC* coke combustion; *COC* crude oil combustion; *FOC* fuel oil combustion; *EPC* electric power combustion; ** $p < 0.01$; * $p < 0.05$).

prove the effectiveness of China's air pollution control policies in recent years⁴³, especially the "Action Plan on Air Pollution Prevention and Control"⁴⁴. A series of air quality improvement measures, such as the "Clear water and blue sky" project and "Strengthening Measures of Air Pollution Prevention and Control in Wuhan," have been implemented in Wuhan city in recent years. These air quality improvement measures focus on controlling automobile exhaust emissions and gradually raising road vehicle exhaust emission standards, accelerating the renovation of coal-fired boilers, and controlling pollution sources, such as dust and volatile organic compounds. However, the annual average O₃_8h_C showed no clear decrease trend in this study, consistent with Chen et al.²³ The temporal variations in air pollutants indicated that Wuhan's air pollution control measures have achieved certain results, but challenges still exist.

Spatial variations in air pollutants. The seasonal patterns of PM_{2.5}_C, PM₁₀_C, SO₂_C, NO₂_C, CO_C and O₃_8h_C were similar at all 22 sites, indicating that the formation processes of air pollutants were not significantly influenced by the locations. However, the pollutants showed small differences between the stations

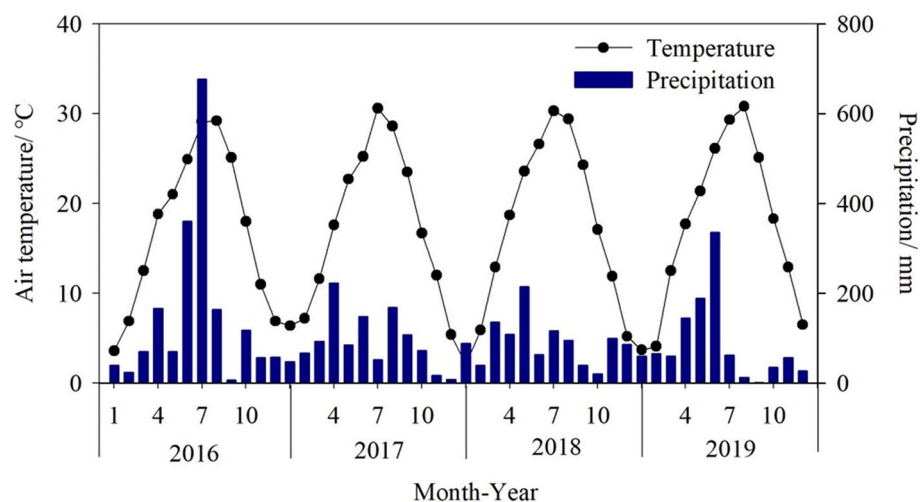


Figure 5. Monthly air temperature and precipitation during 2016–2019 in Wuhan.

in the same year. The $PM_{2.5_C}$ and PM_{10_C} were higher in urban and industrial sites and lower in the control site, indicating high particulate matter pollution at urban and industrial sites. Wang et al.⁴² also found higher $PM_{2.5_C}$ and PM_{10_C} at industrial and urban sites than at mountainous sites. The SO_2_C were higher in industrial sites, suggesting that industrial manufacturing processes were a significant pollution source of SO_2 emission in Wuhan. The findings are consistent with Syaferi et al.¹⁰. The NO_2_C was lower, while $O_3_8h_C$ was higher in suburban sites is consistent with Wang et al.⁴², while the lower NO_2_C in suburban areas was mainly related to the lower traffic volumes there¹⁰. Previous studies have also shown that automobile exhaust is the primary source of urban nitrogen oxide pollution⁴². The annual average CO_C during 2016–2020 showed no spatial differences were not consistent with Syaferi et al.¹⁰ and Wang et al.⁴². This might mainly be attributed to the consistency in the spatial distribution of pollutant emissions⁴⁵ and the impact of land use/land coverage in Wuhan city. Therefore, the extent of development and land use/land coverage should be considered further to explore spatial differences in the concentrations of air pollutants. Thus, urban green space could be integrated, and landscape patterns could be optimized to improve air quality and decrease air pollutant concentrations.

Correlations between air pollutants and influencing indicators. Significant positive correlations between $PM_{2.5_C}$, PM_{10_C} , SO_2_C , NO_2_C , and CO_C suggested that they had originated from the same sources or were impacted by the same drivers^{23,46}. This indicated that control measures could simultaneously decrease the concentrations of these pollutants²³. However, the negative correlations between $O_3_8h_C$ and the other pollutants indicate the difficulty in controlling the six pollutants simultaneously^{46,47}. Further studies should be explored to reveal the formation and control strategies of O_3_8h in Wuhan.

Correlations between the six pollutants and the meteorological indicators suggested that the wet deposition process (e.g., scavenging and wash-out) attributed to the increase in *Prec* could reduce $PM_{2.5_C}$, PM_{10_C} , SO_2_C , NO_2_C , and CO_C ^{48,49}. The rise in air temperature, on the one hand, could intensify the activity of atmospheric molecules to some extent, leading to the diffusion of $PM_{2.5}$, PM_{10} , SO_2 , NO_2 , and CO . On the other hand, the rise in temperature could intensify the photochemical reaction in the atmosphere, leading to the rise of $O_3_8h_C$. This also explains the monthly and seasonal pattern of $PM_{2.5_C}$, PM_{10_C} , SO_2_C , NO_2_C , CO_C , and $O_3_8h_C$ ^{23,50}. There was a significant negative correlation between *RH* and $O_3_8h_C$, indicating that ozone was easy to accumulate under low humidity conditions, and ozone concentration decreased with increasing relative air humidity. Wind speed showed no significant relationship with the air pollutants in this study, showing that wind speed did not intensify air turbulence to improve air quality in Wuhan.

$PM_{2.5_C}$, PM_{10_C} , SO_2_C , NO_2_C , and CO_C had a significantly negative relationship with *GDP*, *PGDP*, *PD*, *PRP*, *RA*, *CV*, and *GRB*. $PM_{2.5_C}$, SO_2_C , and NO_2_C displayed a significantly positive relationship with coal and energy consumption. The increase in *PPGA* and *GRB* and the decrease in *EC*, *CAC*, *CKC*, *COC*, *FOC*, and *EPC* have compensated for the adverse effects of the increase in *GDP*, *PGDP*, *PRP*, *PD*, *RA*, *CV* in the same period, leading to an overall decrease in $PM_{2.5_C}$, PM_{10_C} , SO_2_C , NO_2_C , and CO_C from 2009 to 2020 in Wuhan. An increase in *PPGA* and *GRB* mainly reveals the investments in environmental protection, while a decrease in *EC*, *CAC*, *CKC*, *COC*, *FOC*, and *EPC* mainly reveals the controls on anthropogenic sources, indicating that increase in atmospheric environmental carrying capacity and effective emission-cutting measures were effective in reducing the accumulation of $PM_{2.5}$, PM_{10} , SO_2 , NO_2 , and CO . Air pollutants were inversely proportional to vehicles ownership, which was attributed to the widespread use of clean energy vehicles. Compared to the other pollutants, the correlation between $O_3_8h_C$ and the socioeconomic factors suggested that the control of $O_3_8h_C$ is still a challenge for Wuhan. The urban sites with the highest population densities, high numbers of automobiles, and clustered commercial living zones have relatively stable air pollutant emissions. The air pollution caused by municipal emissions has become a dominant source in urban areas. The industrial site with

high energy consumption intensity, leading to a high concentration of heavy pollution industrial activities. The air pollutants from urban and industrial sites were also transferred to the other sites, which was a major cause of Wuhan's poor air quality.

Conclusion

This study examined the tempo-spatial characteristics of PM_{2.5-C}, PM_{10-C}, SO_{2-C}, NO_{2-C}, CO-C, and O_{3-8h-C}, monitored at 22 stations in Wuhan city from January 2016 to December 2020, and their correlations with the meteorological and socio-economic factors. Based on 5-year data, we found that the maximum values of PM_{2.5-C}, PM_{10-C}, SO_{2-C}, NO_{2-C}, and CO-C were in winter, while O_{3-8h-C} were in summer. The temporal variations in air pollutants showed that annual average PM_{2.5-C}, PM_{10-C}, SO_{2-C}, NO_{2-C}, and CO-C were lower in 2020 than in other years. The spatial variations in air pollutants indicated that the PM_{2.5-C} and PM_{10-C} were higher in urban and industrial sites and lower in the control site. The SO_{2-C} was higher in industrial sites. The NO_{2-C} were lower, and O_{3-8h-C} was higher in suburban sites, while CO showed no spatial differences in their concentrations.

Significant correlations were found between the concentrations of the six air pollutants, except for the relevance between O_{3-8h-C} and PM_{10-C}. The correlations between O_{3-8h-C} and the other pollutants were more complex than those between the other 5 pollutants. PM_{2.5-C}, PM_{10-C}, NO_{2-C}, and CO-C were significantly negatively correlated with temperature and precipitation but insignificantly associated with relative air humidity. SO_{2-C} was significantly negatively associated with temperature, relative air humidity, and precipitation. O_{3-8h-C} was significantly positively associated with temperature while significantly negatively associated with relative air humidity. There was no significant correlation between air pollutants and wind speed. Several socioeconomic factors drove air quality concentrations. The urban and industrial sites with high population densities, number of automobiles, and energy consumption intensity, led to a high concentration of pollutants. An increase in atmospheric environmental carrying capacity and effective emission-cutting measures effectively reduced the accumulation of PM_{2.5}, PM₁₀, SO₂, NO₂, and CO. Overall, this study provided scientific insights that tempo-spatial characteristics and major influencing factors should be taken into account to improve Wuhan's air quality.

Data availability

All data analyzed during this study are included in this published article, and could be obtained upon request from B.J. (email: jbshuibao415@126.com).

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Author contributions

Y.C. and B.J. contributed to the study conception and design. Data collection and analysis were performed by Y.C. The first draft of the manuscript was written by Y.C. and all authors commented on previous versions of the manuscript. All authors read and approved the final manuscript. All authors declare the consent to publish.

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Competing interests

The authors declare no competing interests.

Additional information

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