Ambient Air Pollution and Hospitalizations for Ischemic Stroke: A Time Series Analysis Using a Distributed Lag Nonlinear Model in Chongqing, China

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Short-term exposure to air pollution has been associated with ischemic stroke (IS) hospitalizations, but the evidence of its effects on IS in low- and middle-income countries is limited and inconsistent. We aimed to quantitatively estimate the association between air pollution and hospitalizations for IS in Chongqing, China. This time series study included 2,299 inpatients with IS from three hospitals in Chongqing from January 2015 to December 2016. Generalized linear regression models combined with a distributed lag nonlinear model (DLNM) were used to investigate the impact of air pollution on IS hospitalizations. Stratification analysis was further implemented by sex, age, and season. The maximum lag-specific and cumulative percentage changes of IS were 1.2% (95% CI: 0.4–2.1%, lag 3 day) and 3.6% (95% CI: 0.5–6.7%, lag 05 day) for each 10 µg/m³ increase in PM$_{2.5}$; 1.0% (95% CI: 0.3–1.7%, lag 3 day) and 2.9% (95% CI: 0.6–5.2%, lag 05 day) for each 10 µg/m³ increase in PM$_{10}$; 4.8% (95% CI: 0.1–9.7%, lag 4 day) for each 10 µg/m³ increase in SO$_2$; 2.5% (95% CI: 0.3–4.7%, lag 3 day) and 8.2% (95% CI: 0.9–16.0%, lag 05 day) for each 10 µg/m³ increase in NO$_2$; 0.7% (95% CI: 0.0–1.5%, lag 6 day) for each 10 µg/m³ increase in O$_3$. No effect modifications were detected for sex, age, and season. Our findings suggest that short-term exposure to PM$_{2.5}$, PM$_{10}$, SO$_2$, NO$_2$, and O$_3$ contributes to more IS hospitalizations, which warrant the government to take effective actions in addressing air pollution issues.

Keywords: air pollution, short-term, ischemic stroke, hospitalizations, DLNM

INTRODUCTION

Stroke is a predominant public health concern in the world and the leading cause of death in China, which caused 6.2 million deaths and 132 million disability-adjusted life years globally in 2017 (1, 2). In China, ischemic stroke (IS) is the main subtype of stroke, accounting for 80% of all stroke events (3). Moreover, the incidence of IS has increased substantially in China in recent years (4), which has generated a huge burden on healthcare costs and the economy. Therefore, identification of modifiable risk factors for IS has substantial public health implications.
Many factors have been confirmed to be related to IS, including smoking, lacking of physical exercise, hypertension, obesity, and atrial fibrillation (5). In addition, increasing epidemiological evidence has shown a striking relationship between air pollution exposure and IS (6, 7). A study conducted in nine US counties demonstrated the short-term effects of PM$_{10}$, CO, SO$_2$, and NO$_2$ on hospitalizations for IS (8). Another study by Tian, Y. et al. observed a significant increase in hospitalizations for IS with transient increases of PM$_{2.5}$, SO$_2$, NO$_2$, O$_3$, and CO in China (9). However, most of these studies were conducted in high-income countries, and the scientific evidence that generated in low- or middle-income countries was scarce, especially in China (10). Recently, data from the National Epidemiological Survey of Stroke in China showed that the estimated mortality-to-incidence ratio (MIR) of stroke is the highest in the southwest and the lowest along the eastern and southern coasts; the proportion of registered medical doctors per 1,000 of the population is the highest in northern and eastern China and the lowest in the southwest (4). Furthermore, the geographical regions with high correlation between IS mortality and PM$_{2.5}$ gradually moved from western and northern China to the southwest during 1990–2015. Therefore, it was of great importance to explore the associations between air pollution and IS in the southwest.

Chongqing is a major heavy industrialized city located in Southwest China. Heavy industry and valley basin structure contributed to some of the worst air pollution in comparison with China’s other cities. Thus, it is more meaningful to study the associations between the ambient air pollution and the occurrence of IS in Chongqing.

In this study, we conducted a time series analysis to investigate the associations between short-term exposure to air pollution and daily IS hospitalizations in Chongqing, China.

**MATERIALS AND METHODS**

**Study Area and Health Data Collection**

Chongqing is a major heavy industrialized city and one of the four municipalities in China. It is an urban city with an area of 5472.68 km$^2$ and a population of greater than 8.6 million in 2017 (11).

Data on daily hospital admissions for IS in this study were retrieved from 3 tertiary-level comprehensive hospitals (The Second Affiliated Hospital of Chongqing Medical University, University-Town Hospital of Chongqing Medical University, and The Southeast Hospital) with approximately 3,500, 1,500, and 1,200 inpatient beds, respectively. The medical information was recorded on Platform of Medical Data Science Academy of Chongqing Medical University, and it included the patients’ age, sex, diagnosis, dates of admission, and discharge. We identified admissions for IS (International Classification of Diseases (ICD)-10 code I63) from January 1, 2015 to December 31, 2016 according to the 10th revision of ICD-10 codes.

**Environmental Data**

Data on air pollution, including levels of PM$_{2.5}$, PM < 10$\mu$m in aerodynamic diameter (PM$_{10}$), sulfur dioxide (SO$_2$), nitrogen dioxide (NO$_2$), carbon monoxide (CO), and ozone (O$_3$) between January 1, 2015 and December 31, 2016, were obtained from the China Air Quality Online Monitoring and Analysis Platform. There were 17 fixed monitoring stations that measure concentrations of air pollution in Chongqing. In the Chinese air quality online monitoring system, PM$_{10}$ and PM$_{2.5}$ were monitored by continuous automatic $\beta$-ray monitoring method, SO$_2$ and O$_3$ by ultraviolet fluorescence, and NO$_2$ by chemiluminescence. These measurements were completed in adherence to the China National Quality Control (GB3095-2012) protocol. We derived 24-h mean concentrations of these pollutants in Chongqing to represent each individual’s daily exposure levels except for ozone (by averaging 8-h maximum values). Weather conditions including daily mean temperature and mean relative humidity were sourced from the National Meteorological Information Center of China. All the monitoring stations and hospitals were located in urban areas (Figure 1).

**Statistical Analysis**

Spearman’s correlation coefficient was applied to explore the correlation between air pollutants and meteorological factors. For the time series analysis, we used generalized linear model combined with a distributed lag non-linear model (DLNM) to investigate both non-linear and delayed effects of air pollutants on daily hospitalizations for IS. As daily IS hospital admissions data being quantitative data, belonging to small probability events, and conforming to Poisson distribution, there a Poisson regression models allowing for overdispersion was used. The model is as follows:

\[
\log(Y_t) = \alpha + \beta X_t, \text{lag} + \beta \text{Mean Tem}_t, \text{lag} + ns(\text{RH}_t, df = 3) + ns(\text{Mean Tem}_t, df = 7 \times \text{years}) + \text{factor (DOW)} + \text{factor (holiday)}
\]

where $t$ is day of observation, $Y_t$ refers to the number of IS hospital admissions, $X_t$ (lag) and $\text{Mean Tem}_t$ (lag) indicate the matrix of air pollution and mean temperature by applying the DLNM, $X_t$ and $\text{Mean Tem}_t$ represent pollutant concentrations and mean temperature at day $t$, $ns(\text{RH}_t, df = 3)$ are natural cubic spline with 3 degrees of freedom for daily relative humidity, $ns(\text{Mean Tem}_t, df = 7 \times \text{years})$ is 7 degrees of freedom per year, to control long-term trend and seasonality (12, 13), DOW is the day of the week, and holiday is the variable for the holiday effect. We adopt a linear and a natural cubic spline ($df = 3$) function to fit the exposure–response relationship and the lag–response relationship, respectively (13). Also, the parameter of lag–response relationship between mean temperature and IS was equal to air pollutants. Furthermore, the non-linear relationship between air pollution and IS was explored by a natural cubic spline function ($df = 3$) when the cumulative effect appears to be strongest.

Previous studies had shown that the lagged effect of air pollutants was usually short. In this study, single-day lags (from lag 0 day to lag 7 day) combined with cumulative lags (from lag 01 day to lag 07 day) were applied to evaluate the lagged effects of air pollutants. We initially performed single-pollutant model to assess the association between air pollution and IS
FIGURE 1 | Spatial distributions of air quality monitoring stations, weather monitoring stations, and addresses of hospitals in Chongqing, China.
hospitlizations, and then, the significant air pollutants were included in subsequent analysis.

We conducted stratification analysis to explore the potential effect modification by age (<75, ≥75 years), sex (male, female), and season [cold season, warm season]. The risk estimates were expressed in terms of the percentage changes (excess risk, ER (‰)) and season [cold season, warm season]. The risk estimates were expressed in terms of the percentage changes (excess risk, ER (%) and season [cold season, warm season]. The risk estimates were expressed in terms of the percentage changes (excess risk, ER (%) and season [cold season, warm season]. The risk estimates were expressed in terms of the percentage changes (excess risk, ER (%) and season [cold season, warm season]. The risk estimates were expressed in terms of the percentage changes (excess risk, ER (%)

| Variables                      | Mean±SD | Min | P10 | P25 | P50 | P75 | P90 | P95 | P99 | Max |
|-------------------------------|---------|-----|-----|-----|-----|-----|-----|-----|-----|-----|
| Cardiovascular and cerebrovascular disease | 20.48 ± 9.49 | 2.0 | 9.0 | 13.0 | 19.0 | 26.0 | 34.0 | 38.0 | 47.0 | 57.0 |
| IS                           | 3.15 ± 2.00 | 0.0 | 1.0 | 2.0 | 3.0 | 4.0 | 6.0 | 7.0 | 9.0 | 13.0 |
| Gender                       |         |     |     |     |     |     |     |     |     |     |
| Male                         | 1.35 ± 1.23 | 0.0 | 0.0 | 0.0 | 1.0 | 2.0 | 3.0 | 4.0 | 5.0 | 7.0 |
| Female                       | 1.80 ± 1.45 | 0.0 | 0.0 | 1.0 | 2.0 | 3.0 | 4.0 | 5.0 | 6.0 | 7.0 |
| Age                          |         |     |     |     |     |     |     |     |     |     |
| <75                          | 1.07 ± 1.08 | 0.0 | 0.0 | 0.0 | 1.0 | 2.0 | 3.0 | 3.0 | 4.0 | 6.0 |
| ≥75                          | 2.07 ± 1.55 | 0.0 | 0.0 | 1.0 | 2.0 | 3.0 | 4.0 | 5.0 | 6.0 | 10.0 |
| Season                       |         |     |     |     |     |     |     |     |     |     |
| Warm                         | 3.06 ± 1.95 | 0.0 | 1.0 | 2.0 | 3.0 | 4.0 | 6.0 | 7.0 | 8.0 | 13.0 |
| Cold                         | 3.23 ± 2.04 | 0.0 | 1.0 | 2.0 | 3.0 | 5.0 | 6.0 | 7.0 | 9.0 | 10.0 |
| PM2.5 (µg/m³)                | 54.41 ± 30.72 | 10.0 | 25.0 | 34.0 | 46.0 | 66.0 | 93.0 | 114.0 | 164.1 | 212.0 |
| PM10 (µg/m³)                 | 81.75 ± 40.68 | 13.0 | 41.0 | 54.0 | 74.0 | 98.0 | 131.0 | 164.5 | 228.8 | 293.0 |
| SO₂ (µg/m³)                  | 14.55 ± 6.59 | 4.0 | 7.0 | 10.0 | 13.0 | 18.0 | 23.0 | 27.0 | 36.4 | 42.0 |
| CO (mg/m³)                   | 1.04 ± 0.28 | 0.4 | 0.7 | 0.9 | 1.0 | 1.2 | 1.4 | 1.5 | 3.4 |     |
| NO₂ (µg/m³)                  | 45.05 ± 12.69 | 16.0 | 30.0 | 36.0 | 44.0 | 53.0 | 62.0 | 69.0 | 79.0 | 96.0 |
| O₃ (µg/m³)                   | 68.11 ± 45.38 | 4.0 | 15.0 | 29.0 | 60.0 | 101.0 | 135.0 | 184.3 | 223.0 |     |
| Temperature (°C)             | 19.60 ± 7.47 | 1.2 | 9.6 | 13.0 | 20.1 | 25.2 | 29.4 | 31.4 | 34.3 | 36.2 |
| Relative humidity (%)         | 75.96 ± 10.93 | 43.0 | 60.3 | 68.0 | 76.0 | 84.0 | 90.0 | 92.0 | 94.6 | 96.3 |

PM2.5: particles with aerodynamic diameter < 2.5 µm; PM10: particles with aerodynamic diameter < 10 µm; SO₂: sulfur dioxide; NO₂: nitrogen dioxide; CO: carbon monoxide; O₃: ozone; IS: ischemic stroke.

Table 1 shows the distribution of daily IS admissions, air pollutants, and meteorological factors in Chongqing, China (January 2015–December 2016). The data were collected from 3 hospitals during January 1, 2015 to December 31, 2016. Overall, 2,299 patients with IS formed the basis of this study, with a daily average of 3 cases. 57.1% were men and 65.9% were women aged ≥75 years. The mean 24-h PM2.5, PM10, SO₂, CO, NO₂, and O₃ concentrations were 54.41 µg/m³, 81.75 µg/m³, 14.55 µg/m³, 1.04 mg/m³, 45.05 µg/m³, and 68.11 µg/m³, respectively.

The Spearman’s correlation coefficients between air pollutants and meteorological factors are listed in Table 2. PM2.5, PM10, SO₂, NO₂, and CO were positively and moderately or strongly correlated. O₃ exposure was negatively and weakly associated with PM2.5, SO₂, and CO, whereas the correlation coefficient with PM10 and NO₂ was not statistically significant. Temperature was positively associated with O₃, and negatively associated with other air pollutants, whereas relative humidity was positively associated with CO, and negatively associated with other air pollutants.

Figure 2 shows that exposure to PM2.5, PM10, SO₂, NO₂, and O₃ was significantly associated with IS. In single-day lag structures, exposures to PM2.5 (lag 2 day–lag 4 day), PM10 (lag 2 day–lag 5 day), SO₂ (lag 4 day), NO₂ (lag 2 day, lag 3 day), and O₃ (lag 6 day) were associated with increased IS cases. The maximum lag-specific percentage changes for each 10 µg/m³ increase in PM2.5, PM10, SO₂, NO₂, and O₃ were 1.7% (95% CI: 0.4–2.1%, lag 3 day), 1.0% (95% CI: 0.3–1.7%, lag 3 day), 4.8% (95% CI: 0.1–9.7%, lag 4 day), 2.5% (95% CI: 0.3–4.7%, lag 3 day), and 0.7% (95% CI: 0.0–1.5%, lag 6 day), respectively. In cumulative lag structures, the unfavorable effects of PM2.5, PM10, and NO₂ occurred to lag 04 day–lag 06 day, lag 04 day–lag 06 day, and lag 04 day–lag 05 day, and the corresponding maximum percentage

RESULTS

Basic characteristics of patients with IS, meteorological factors, and air pollution are displayed in Table 1. There was a total of 14,969 hospital admissions for cardiovascular and cerebrovascular diseases from 3 hospitals during January 1, 2015 to December 31, 2016. Overall, 2,299 patients with IS formed the basis of this study, with a daily average of 3 cases. 57.1% were men and 65.9% were women aged ≥75 years. The mean 24-h PM2.5, PM10, SO₂, CO, NO₂, and O₃ concentrations were 54.41 µg/m³, 81.75 µg/m³, 14.55 µg/m³, 1.04 mg/m³, 45.05 µg/m³, and 68.11 µg/m³, respectively.

The Spearman’s correlation coefficients between air pollutants and meteorological factors are listed in Table 2. PM2.5, PM10, SO₂, NO₂, and CO were positively and moderately or strongly correlated. O₃ exposure was negatively and weakly associated with PM2.5, SO₂, and CO, whereas the correlation coefficient with PM10 and NO₂ was not statistically significant. Temperature was positively associated with O₃, and negatively associated with other air pollutants, whereas relative humidity was positively associated with CO, and negatively associated with other air pollutants.

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TABLE 2 | Spearman’s correlation coefficients between air pollutants and meteorological factors.

| Variables | PM$_{2.5}$ | PM$_{10}$ | SO$_2$ | CO | NO$_2$ | O$_3$ | Temperature | Relative humidity |
|-----------|------------|-----------|--------|----|--------|------|-------------|--------------------|
| PM$_{2.5}$ | 1.00       |           |        |    |        |      |             |                    |
| PM$_{10}$  | 0.95**     | 1.00      |        |    |        |      |             |                    |
| SO$_2$     | 0.61**     | 0.68**    | 1.00   |    |        |      |             |                    |
| CO         | 0.60**     | 0.56**    | 0.56** | 1.00|        |      |             |                    |
| NO$_2$     | 0.71**     | 0.74**    | 0.61** | 0.53** | 1.00  |      |             |                    |
| O$_3$      | −0.09*     | 0.04      | −0.15**| −0.44**| −0.04 | 1.00 |             |                    |
| Temperature| −0.26**    | −0.15**   | −0.35**| −0.41**| −0.23**| 0.73**| 1.00       |                    |
| Relative humidity | −0.12** | −0.29**   | −0.39**| 0.23** | −0.14**| −0.65**| −0.38** | 1.00             |

PM$_{2.5}$, particles with aerodynamic diameter < 2.5 μm; PM$_{10}$, particles with aerodynamic diameter < 10 μm; SO$_2$, sulfur dioxide; NO$_2$, nitrogen dioxide; CO, carbon monoxide; O$_3$, ozone; * p < 0.05; ** p < 0.01.

FIGURE 2 | Percentage changes [excess risk, ER (%)] of IS cases for each pollutant at various lags.

changes were 3.6% (95% CI: 0.5–6.7%, lag 05 day), 2.9% (95% CI: 0.6–5.2%, lag 05 day), 8.2% (95% CI: 0.9–16.0%, lag 05 day) (Supplementary Table S1), whereas no association between CO and IS had been detected.

The exposure–response curves for PM$_{2.5}$, PM$_{10}$, SO$_2$, NO$_2$, CO, and O$_3$ with hospitalizations for IS are displayed in Figure 3. The exposure–response curve for NO$_2$, SO$_2$, and O$_3$ showed thresholds for their associations with hospitalizations for IS: 50–75 μg/m$^3$ for NO$_2$, 20–30 μg/m$^3$ for SO$_2$, and 100–150 μg/m$^3$ for O$_3$. The exposure–response relationship for PM$_{2.5}$, PM$_{10}$, and CO was almost linear.

Stratified and sensitivity analyses were performed on the basis of reaching the maximum effect of PM$_{2.5}$, PM$_{10}$, SO$_2$, NO$_2$, and O$_3$ in the cumulative lag structure. The associations
between air pollutant exposures and the hospitalizations for IS were evaluated in subgroups based on sex, age, and season (Table 3). For PM$_{2.5}$, PM$_{10}$, SO$_2$, and NO$_2$, the effects were still significant in cold season subgroup. In addition, both sex and the younger received adverse effects when exposed to PM$_{10}$, whereas NO$_2$ seems to be more susceptible to men and the elder. Although the percentage changes were somewhat different and some of them were insignificant, we did not find any significant effect modification across sex, age, and season (all $p$ for effect modification $> 0.05$).

Table 4 provides the results of the two-pollutant model, and further adjustment for other pollutant exposures did not materially change the associations between PM$_{2.5}$, PM$_{10}$, SO$_2$, NO$_2$, and O$_3$ exposures and IS. Results of the sensitivity analysis (Supplementary Table S2) indicated that the effect estimates of the association between air pollution and IS hospitalizations were not substantially affected with the use of alternative $d_f$ value for time (6–8 per year), temperature (3–6), and relative humidity (3–6), although some of the associations became insignificant for temperature ($d_f = 6$).

**DISCUSSION**

There is growing interest in the associations between ambient air pollution and risk of IS, but data on the associations in low-
or middle-income cities are still limited, especially in Southwest China. Chongqing is mainly associated with heavy industry with a valley basin structure, which has a great impact on ambient air quality. Therefore, we conducted this time series study combining with DLNM to evaluate the short-term effects of ambient air pollutants on hospitalizations for IS in Chongqing. The data obtained in this analysis indicated that short-term exposure to PM$_{2.5}$, PM$_{10}$, NO$_2$, O$_3$, and SO$_2$ was significantly associated with increased hospitalizations for IS. Age, sex, and season did not appear to significantly modify the associations between short-term exposure to air pollution and IS onset. The sensitivity analysis showed a robustness of the pollutant model results.

In single-pollutant models, we found that each 10 µg/m$^3$ increase in PM$_{2.5}$, PM$_{10}$, SO$_2$, NO$_2$, and O$_3$ was associated with a 1.2, 1.0, 4.8, 2.5, and 0.7% increase in IS hospitalizations, respectively, which were similar to several previous studies (14–16). For example, Hu’s study conducted in Yancheng city has shown that IS hospitalizations increased 1.06% (95% CI: 0.21–1.91%) per 10 µg/m$^3$ increase in PM$_{2.5}$ (14). Another multicity study reported that per 10 µg/m$^3$ increase in PM$_{2.5}$, PM$_{10}$, NO$_2$, and SO$_2$ remained significant in cold season. This may be attributable to a high number of foggy days during cold season in Chongqing. During our study period, the mean AQI in Chongqing (80.75) was higher than a less foggy city, such as Guangzhou (68.79), which was similar to Chongqing in population and land area (24).

The biological mechanism of air pollution-induced IS has not been fully investigated. Several possible mechanisms have been proposed to be related to the inflammation (25), oxidative stress (26), abnormal lipid metabolism (7, 27), and autonomic dysfunction (7). For example, PM$_{2.5}$ can trigger the release of pro-inflammatory mediators causing systemic inflammation, which impaired blood–brain barrier (BBB) stability and induces to the generation of ROS and oxidative stress (28). Inhalation of SO$_2$ can affect heart rate variability, increase oxidation, and exacerbate blood clotting and thrombosis formation (29). In addition, a study showed that exposure to O$_3$ and particulate pollutants increased endothelin 1, leading to vascular endothelial dysfunction and subsequent brain damage (30).

Our study has some potential limitations. First, small sample sizes might lead to lower statistical power; therefore, subsequent studies with large sample sizes would be expected. Second, city-level concentration of air pollution rather than individual exposure was utilized as the exposure concentration, which might underestimate the effect of air pollution. Third, we selected only 3 hospitals in one single city, so the generalization of the results requires caution. Finally, although we adjusted for several confounders such as seasonality, day of week, public.
holiday, and weather conditions, there might exist some other confounding factors.

CONCLUSIONS

This study suggests that short-term exposure to PM$_{2.5}$, PM$_{10}$, NO$_2$, O$_3$, and SO$_2$ was associated with increased hospitalizations for IS in Chongqing, China. Our study provides new evidence on the association between air pollution and IS. Further studies are warranted to help government effectively reduce the burden of air pollution.

DATA AVAILABILITY STATEMENT

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

ETHICS STATEMENT

The studies involving human participants were reviewed and approved by Ethics Committee of Chongqing Medical University in Chongqing, China. Written informed consent to participate in this study was provided by the participants’ legal guardian/next of kin.

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AUTHOR CONTRIBUTIONS

HC wrote the manuscript and analyzed the data. PL and YD collected and inputted the data. JF, YX, ML, ZC, KP, and LZ reviewed the results and provided guidelines for presentation and interpretation. All authors have read and approved the final manuscript.

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SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fpubh.2021.762597/full#supplementary-material
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