### **RESEARCH ARTICLE**



# A time-series analysis of short-term ambient ozone exposure and hospitalizations from acute myocardial infarction in Henan, China

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#### **Abstract**

Epidemiological studies in recent years have identified an association between exposure to air pollutants and acute myocardial infarction (AMI); however, the association between short-term ozone ( $O_3$ ) exposure and AMI hospitalization remains unclear, particularly in developing countries. Therefore, this study collected information on 24,489 AMI patients, including daily air pollutant and meteorological data in Henan, China, between 2016 and 2021. A distributed lagged nonlinear model combined with a Poisson regression model was used to estimate the nonlinear lagged effect of  $O_3$  on AMI hospitalizations. We also quantified the effects of  $O_3$  on the number of AMI hospitalizations, hospitalization days, and hospitalization costs. The results showed that single- and dual-pollution models of  $O_3$  at lag0, lag1, and lag (01–07) were risk factors for AMI hospitalizations, with the most significant effect at lag03 (RR=1.132, 95% CI:1.083–1.182). Further studies showed that males, younger people (15–64 years), warm seasons, and long sunshine duration were more susceptible to  $O_3$ . Hospitalizations attributable to  $O_3$  during the study period accounted for 11.66% of the total hospitalizations, corresponding to 2856 patients, 33,492 hospital days, and 90 million RMB. Maintaining  $O_3$  at 10–130 µg/m<sup>3</sup> can prevent hundreds of AMI hospitalizations and save millions of RMB per year in Henan, China. In conclusion, we found that short-term exposure to  $O_3$  was significantly associated with an increased risk of hospitalization for AMI in Henan, China, and that further reductions in ambient  $O_3$  levels may have substantial health and economic benefits for patients and local healthcare facilities.

Keywords Acute myocardial infarction · Ozone · Air pollution · Distributed lag nonlinear model · Time-series analysis

### Introduction

Acute myocardial infarction (AMI) is known for its high morbidity and mortality (Hoole and Bambrough 2020, Reed et al. 2017); in China, AMI mortality has more than doubled in the past 20 years (Yang et al. 2013). The World Bank estimates that the number of patients with AMI in China will

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Yulong Wei and Lin Fei contributed equally to this work.

#### Highlights

- Association between air pollutants exposure and acute myocardial infarction exists.
- Determining lagged effect of O<sub>3</sub> on acute myocardial infarction using two models.
- $\bullet$  Hospitalizations caused by  ${\rm O}_3$  accounted for 11.66% of the total hospitalizations.

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reach 23 million by 2030 (Bank. 2013). There is increasing evidence that exposure to ambient air pollutants is associated with an increased incidence of AMI (Chen et al. 2018; Ishii et al. 2021; Liu et al. 2021). However, most of these are studies on particulate matter and AMI incidence, with few involving ozone (O<sub>3</sub>), particularly in developing countries, and inconsistent outcomes. O<sub>3</sub> is a secondary pollutant with strong oxidizing properties (Liu et al. 2022). The consequent temperature increase may lead to increased O<sub>3</sub> concentrations in the future owing to the increased greenhouse gas emissions in the environment, accelerating global warming (Pu et al. 2017). Wang et al. (2017) reported that this process may further accelerate with the rapid consumption of fossil fuels in China. The World Health Organization (WHO) updated the WHO Global Air Quality Guidelines in 2021 to address the health hazards of O<sub>3</sub> exposure by adding a new target of ambient seasonal O<sub>3</sub> concentrations of less than 60 μg/m<sup>3</sup> (Burki 2021); thus, ambient O<sub>3</sub> pollution is a serious global public health problem.

In environmental ecology studies, short-term exposure to O<sub>3</sub> increases the risk of cardiovascular disease, stroke, respiratory disease onset, and mortality (EPA US 2020). Recently, large cohort studies in the USA and Canada have shown that long-term exposure to  $O_3$  may be associated with mortality from cardiovascular diseases (Cakmak et al. 2016; Lim et al. 2019). In addition, a meta-study reported evidence of increased all-cause and cardiovascular mortalities associated with exposure to high O<sub>3</sub> levels (Stafoggia et al. 2010). However, most of these studies were conducted at longterm O<sub>3</sub> exposure levels and in developed countries where O<sub>3</sub> concentrations are mostly below 60 μg/m<sup>3</sup> (the ambient air quality standard for O<sub>3</sub> in China is 160 μg/m<sup>3</sup> (Guan et al. 2016)). In contrast, few studies have been conducted on short-term O<sub>3</sub> exposure; no studies have analyzed the relationship between short-term O<sub>3</sub> exposure and AMI hospitalization in China. Therefore, we conducted a multicenter time-series study in Henan Province, China, to explore the effect of short-term O<sub>3</sub> exposure on AMI hospitalization. In addition, this study quantified the fraction of hospitalizations attributable to O<sub>3</sub>; the number of AMI-related hospitalizations and hospitalization costs saved by falling below China's national secondary air quality standards was quantified to make appropriate recommendations for relevant public health policies.

### **Materials and methods**

# Study population

In this study, data were obtained from seven large general hospitals (Table S1) in Henan Province, China (31°23′–36°22′N, 110°21′–116°39′E), located in five cities: Anyang, Xinxiang, Luoyang, Kaifeng, and Nanyang (Figure S1). The coordinates of the monitoring stations and hospitals were determined on regional maps using ArcGIS (version 10.5). The selected hospitals were ranked as the top tertiary hospitals in prefecture-level cities and were located in densely populated areas with convenient transportation. With advanced equipment, complete medical departments, and strong diagnosis and treatment capabilities, they have a reputation for diagnosing and treating cardiovascular diseases. Therefore, they reflect the incidence of common cardiovascular diseases in the local area (Wang et al. 2020a). Between January 1, 2016, and October 31, 2021, 24,489 patients with AMI in the study sites mentioned above were enrolled in this study; all acute myocardial infarctions were clinically diagnosed and validated (Thygesen et al. 2018) with diagnosis code I21, from the 10th edition of the International Classification of Diseases (ICD-10). We collected information on these inpatients, including hospitalization number, inpatient medical records, sex, age, home address, and

first diagnosis information. Each patient's hospital address was geocoded to longitude and latitude and matched to the nearest monitoring station (Table S2). These data were also categorized according to sex, age, season, and sunshine duration. The age groups were as follows: younger (15–64 years) and older adults ( $\geq$ 65 years). The cold season was from November to April (next year), and the warm season was from May to October. We set the short ( $\leq$ 6.3 h) and long sunshine duration groups (>6.3 h) based on the average sunshine duration in the warm season (6.33 h).

# Air pollution and weather data

Air pollutant concentrations in the study area:  $O_3$ , particulate matter with aerodynamic matter  $\leq 2.5 \, \mu m \, (PM_{2.5})$ , particulate matter with aerodynamic matter  $\leq 10 \, \mu m \, (PM_{10})$ , carbon monoxide (CO), nitrogen dioxide (NO<sub>2</sub>), and sulfur dioxide (SO<sub>2</sub>) from the China General Environmental Monitoring Station (http://www.cnemc.cn/). Air pollution data were collected from the 28 fixed-site monitoring stations mentioned above, located away from emission sources, such as roads, industries, and factories, and whose measurements reflect overall city averages rather than local pollution sources. For PM<sub>2.5</sub>, PM<sub>10</sub>, CO, NO<sub>2</sub>, and SO<sub>2</sub>, the daily concentrations are 24-h averages, while the diurnal concentrations of O<sub>3</sub> vary significantly; therefore, the maximum 8-h averages of all effective monitoring stations were selected (Liu et al. 2022).

Meteorological data were obtained from the China National Meteorological Information Center (http://data.cma.cn/). Two meteorological parameters (daily average relative humidity and daily average temperature) were introduced as covariates in the model to adjust for confounding factors. This study was approved by the ethical review committee of The First Affiliated Hospital of Xinxiang Medical University (No. 2020097; Xinxiang, China).

# Statistical analysis

# Modeling the association between O<sub>3</sub> and AMI hospitalizations and subgroup analysis

Daily AMI hospitalizations, O<sub>3</sub>, and meteorological data were linked by admission date for time-series analysis. The number of admissions of patients with AMI followed a Poisson distribution. A proposed Poisson regression model based on a generalized summation linear model and a distributed lag nonlinear model (DLMN) was used to evaluate the association between air pollutants and AMI hospitalizations (Chang et al. 2022). The association between AMI hospitalizations and O<sub>3</sub> was fitted using the "DLMN" analysis package in R software (version 4.12). This was combined with the Akaike information criterion



(Dziak et al. 2020). We included the following covariates in the model: a natural cubic spline (NS) function with 2 degrees of freedom (df) per year was used to control for long-term trends and seasonality; the effects of humidity and temperature were controlled with 2 and 5 df NS functions, respectively; the day-of-week effect (Dow) was set as a covariate; and the Chinese holiday effect was also considered in the model to eliminate the effect of public holidays.

The algebraic expression of the model is as follows:

$$Log(E(Y_t)) = \alpha + \beta X_{t,l} + ns(times, 2) + ns(tum, 2) + ns(temp, 5) + Dow + holidays$$

where  $E(Y_t)$  is the number of AMI hospitalizations on day t,  $\alpha$  is the intercept, and  $\beta X_{t,l}$  is the cross-basis matrix for air pollution. ns() represents the NS function.

To explore the lag association between AMI hospitalizations and  $O_3$ , we set the pollutant as a crossmatrix and used a polynomial function for the lags. We set the lag days to 7 days (Raza et al. 2018; Stafoggia et al. 2010) and rated the lagged cumulative effect of each 10  $\mu$ g/m³ increase in airborne  $O_3$  concentration using single-day lag (lag0-lag7) and multi-day cumulative lags (lag01-lag07). In addition, we stratified the study by sex (male and female), age (young and older patients), season (warm and cool seasons), and warm season sunshine duration (long and short sunshine durations) to identify potentially susceptible subgroups for  $O_3$  exposure.

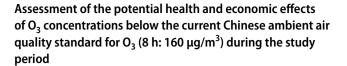
# Calculation of AMI hospitalizations, hospitalization days, and hospitalization costs attributable to $\rm O_3$

Based on the attribution fraction calculation method in previous studies (Chen et al. 2017a; Cheng et al. 2017; Deubner et al. 1975), we estimated the total attribution fraction (AF) and the number of attributions (AN) for AMI hospitalizations. Therefore, the formulae are as follows:

$$AF = 1 - \exp(-\beta) \tag{1}$$

$$AN_i = AF_i * N_i \tag{2}$$

where AF is the attribution fraction for  $O_3$ ,  $AN_i$  is the number of attributions on day i, and  $N_i$  is the total number of AMI hospitalizations, hospitalization days, and hospitalization costs. According to Steenland and Armstrong (2006), formulae (1) can be simplified to AF = (RR - 1)/RR, RR is the cumulative maximum relative risk.



Daily O<sub>3</sub> concentrations below the secondary standard for ambient air quality in China (8 h: 160 µg/m<sup>3</sup>) accounted for 84.2% (1795/2131) of the total study period. We maintained the O<sub>3</sub> concentrations in the range of 10–160 µg/m<sup>3</sup> (every 30 unit increment). The potential health and economic effects were estimated based on five  $O_3$  concentrations (10-40-70-100-130  $\mu g/m^3$ ). The magnitude of the health benefit (i.e., the number of preventable cases per year) was measured as the difference between the annual number of attributed cases at the O<sub>3</sub> concentration level of the Chinese ambient air quality secondary standard and the number of cases below the five O<sub>3</sub> concentration limits developed. Similarly, the magnitude of the economic benefit, that is, the annual hospitalization cost savings (in RMB), is measured as the difference between the annual attributable hospitalization cost at the O<sub>3</sub> concentration level of the Chinese ambient air quality secondary standard and the hospitalization cost below the five established O<sub>3</sub> concentration limits.

We used a cumulative lag of 3 days for risk assessment in constructing the dual-pollution model, subgroup analysis, and assessment of population attribution because  $O_3$  produced the most significant effect estimate at a cumulative lag of 3 days.

#### Sensitivity analysis

First, we constructed a dual-pollution model to test the effect of using the lag of the co-pollution model by adjusting for common pollutants (PM<sub>2.5</sub>, PM<sub>10</sub>, CO, NO<sub>2</sub>, and SO<sub>2</sub>) in the model. Second, the stability of the results was tested by varying the df values (3, 4, 5, 6, and 7) of the long-term trends.

The statistical analysis for this study was performed using the "dlnm" package in R software (version 4.12). This function reports the effect of every  $10 \mu g/m^3$  increase in  $O_3$  on the daily AMI hospitalizations, estimated as relative risk (RR) and 95% confidence interval (95% CI). A two-sided test p-value < 0.05 was considered statistically significant.

# Results

# Study patients and exposure characteristics

Table 1 shows the descriptive statistics for daily AMI hospitalizations, hospitalization days, total hospitalization costs,



meteorological factors, and air pollution during the study period. Between January 1, 2016, and October 31, 2021, we recorded 24,489 patients diagnosed with AMI (ICD10: I21) in these hospitals, with an average of 11.49 admissions daily. The patients were predominantly male (8.35 cases daily) and aged 15–64 years (6.18 cases daily). The daily maximum 8-h average  $O_3$  concentration during the study period was  $104.29 \,\mu\text{g/m}^3$  (range:4–277  $\mu\text{g/m}^3$ ).

Figure 1 shows the time series patterns of daily AMI hospitalizations, O<sub>3</sub> concentrations, and meteorological indicators between 2016 and 2021. The number of AMI admissions showed seasonal characteristics of high in winter and low in summer, with an overall increasing trend between 2016 and 2021. O<sub>3</sub>, average daily temperature, and sunshine duration showed clear seasonal patterns of variation. Figure S2 shows the interval percentages of O<sub>3</sub> concentrations and monthly O<sub>3</sub> trends in the different cities during the study period. Table S3 shows the Spearman correlations between all pollutant indicators and some meteorological indicators, indicating a significant positive correlation between O<sub>3</sub> and temperature (r=0.78), the same positive correlation with sunshine duration (r = 0.54), and a negative correlation between  $O_3$  and the remaining air pollution indicators.

# Effects of O<sub>3</sub> in single- and dual-pollutant models

Table S4 shows the association between different air pollutants of  $O_3$ ,  $PM_{2.5}$ ,  $PM_{10}$ , CO,  $NO_2$ , and  $SO_2$  and AMI hospitalization at different lag days, with short-term  $O_3$  exposure being the most significantly associated with increased risk of AMI hospitalization in this study area compared to other air pollutants. Table 2 and Figure S3 show the single- and dual-pollution models of the association between  $O_3$  on different lag days and AMI hospitalization. We observed that lag0, lag1, and lag (01–07) were all risk factors for AMI hospitalization in the single- and dual-pollution models, with statistically significant differences and the largest effect at lag03 (RR = 1.132, 95% CI:1.083–1.182).

Figure 2 shows the results of the dual-pollution model at lag03, which are insignificantly different from those of the  $O_3$  single-pollution model.

# Associations between $O_3$ and AMI in different subgroups

Figure 3 reports the RR (95% CI) values of O<sub>3</sub>-related AMI hospitalizations by sex, age, season, and different sunshine hours in the warm season. The mean sunshine duration in

the warm season is shown in Table S6 as 6.33 h; we defined warm season sunshine duration  $\leq$  6.3 has short sunshine duration (SSD) and > 6.3 has long sunshine duration (LSD) (Table S7). Among the different subgroups, AMI hospitalizations in the male, younger patients, warm season, and LSD groups were more susceptible to O<sub>3</sub> effects.

In the sex stratification, the largest effect of O<sub>3</sub> on AMI hospitalizations was estimated in lag03 (RR = 1.117, 95% CI:1.058–1.180) for males and lag02 (RR = 1.104, 95% CI:1.028-1.185) for females. In age stratification, the largest effect of O<sub>3</sub> on AMI hospitalizations was estimated in lag02 (RR = 1.154, 95% CI:1.095-1.215) for younger patients and lag03 (RR = 1.109, 95% CI:1.054-1.176) for older adult patients. In seasonal stratification, the largest effect value of O<sub>3</sub> on AMI hospitalizations during the warm season was estimated at lag03 (RR = 1.190, 95% CI:1.124-1.259); the difference was not statistically significant in the cold season. In the stratification of different daylight hours in the warm season, the largest effect value of O<sub>3</sub> on AMI hospitalizations in the LSD group was estimated at lag02 (RR = 1.201, 95% CI:1.077–1.339) and in the SSD group at lag06 (RR = 1.163, 95% CI:1.015–1.286) (Table S5).

Figure 4 shows the exposure–response curves between  $O_3$  and AMI hospitalizations in different subgroups at lag03. Except for the curves in the cold season, which were not statistically significant, the  $O_3$  curves showed a linear trend in all subgroups, and the cumulative RR increased with increasing  $O_3$  concentration.

# Attributable burden of AMI hospitalizations due to O<sub>3</sub>

Table 3 shows the total attributable fraction (AF) and attributable number (AN) of AMI and different subgroups of hospitalizations attributable to  $O_3$ . We estimated that  $O_3$  accounted for 11.66% of the total cases (2,856), 33,492 days of total hospitalization, and 90 million RMB of the total hospitalization costs. Among the different subgroups, the AF and AN were higher in males, younger patients, and warm seasons. When we adjusted for the dual air pollution factors, the AF decreased slightly, as shown in Table S8.

Figure S4 illustrates that if  $O_3$  concentrations could be reduced to levels below current Chinese ambient air quality standards during the study, such as maintaining 130 µg/m³, 185 AMI cases could be prevented yearly, resulting in savings of 6 million RMB. If  $O_3$  were maintained at even lower levels (below 10 µg/m³), 1056 AMI hospitalizations could be prevented yearly, saving more than 30 million RMB and yielding substantial health and economic benefits. These results varied significantly across the subgroups.



**Table 1** Descriptive statistics of daily AMI hospitalizations, air pollution, and meteorological variables in Henan, China, during 2016–2021

	Mean ± SD	Min	25 <sup>th</sup>	50 <sup>th</sup>	75 <sup>th</sup>	Max
Hospitalizations (counts)						
Total	$11.49 \pm 4.61$	1.00	8.00	11.00	14.00	36.00
Males	$8.35 \pm 3.67$	0.00	6.00	8.00	10.00	30.00
Females	$3.14 \pm 2.02$	0.00	2.00	3.00	4.00	14.00
Young	$6.18 \pm 2.89$	0.00	4.00	6.00	8.00	24.00
Elderly	$5.31 \pm 2.90$	0.00	3.00	5.00	7.00	18.00
Hospitalizations (stays in days	)					
Total	$134.79 \pm 55.90$	4.00	96.00	128.00	166.00	441.00
Males	$97.42 \pm 44.82$	0.00	65.00	92.00	124.00	375.00
Females	$37.37 \pm 27.61$	0.00	17.00	32.00	52.00	233.00
Young	$72.10 \pm 36.03$	0.00	46.00	67.00	94.00	262.00
Elderly	$62.67 \pm 37.34$	0.00	36.00	57.00	83.00	262.00
Hospitalizations (costs in 1000	RMB)					
Total	$357.84 \pm 163.79$	8.15	243.85	334.53	450.26	1234.26
Males	$268.2 \pm 135.46$	0.00	170.28	251.86	343.79	888.85
Females	$89.65 \pm 73.46$	0.00	37.50	74.52	127.87	943.94
Young	$205.46 \pm 113.24$	0.00	123.97	189.26	271.56	822.75
Elderly	$152.32 \pm 101.53$	0.00	77.46	136.30	206.27	988.01
Meteorological factors						
Average temperature (°C)	$16.06 \pm 9.78$	-10.20	7.60	16.90	25.00	34.60
Relative humidity (%)	$61.61 \pm 19.20$	9.00	47.00	63.00	77.00	100.00
Sunshine duration (h)	$5.46 \pm 3.93$	0.00	1.40	6.00	8.80	13.50
Air pollutants						
$O_3 (\mu g/m^3)$	$104.29 \pm 52.44$	4.00	62.00	99.00	142.00	277.00
$PM_{2.5} (\mu g/m^3)$	$62.01 \pm 50.89$	3.00	30.00	45.00	76.00	665.00
$PM_{10} (\mu g/m^3)$	$110.05 \pm 72.06$	6.00	63.00	92.00	135.00	915.00
$CO (mg/m^3)$	$1.18 \pm 0.69$	0.20	0.80	1.00	1.40	10.20
$NO_2 (\mu g/m^3)$	$36.53 \pm 18.27$	5.00	22.00	33.00	48.00	168.00
$SO_2 (\mu g/m^3)$	$17.97 \pm 16.6$	2.00	8.00	13.00	22.00	176.00

AMI acute myocardial infarction; young: 15–64 years; elderly: 65+years;  $O_3$  ozone;  $PM_{2.5}$  particulate matter with aerodynamic matter  $\leq$  2.5  $\mu$ m;  $PM_{10}$  particulate matter with aerodynamic matter  $\leq$  10  $\mu$ m; CO carbon monoxide;  $NO_2$  nitrogen dioxide;  $SO_2$  sulfur dioxide; RMB Chinese dollar, 1000 RMB  $\approx$  147 USD in January 2023.

# Sensitivity analyses

The sensitivity analysis results are reported in Table S9; we varied the df range of the long-term trend (3, 4, 5, 6, 7) between total AMI hospitalizations, sex, age, and different seasons and found similar results to our main analysis, indicating that the outcome estimates are robust.

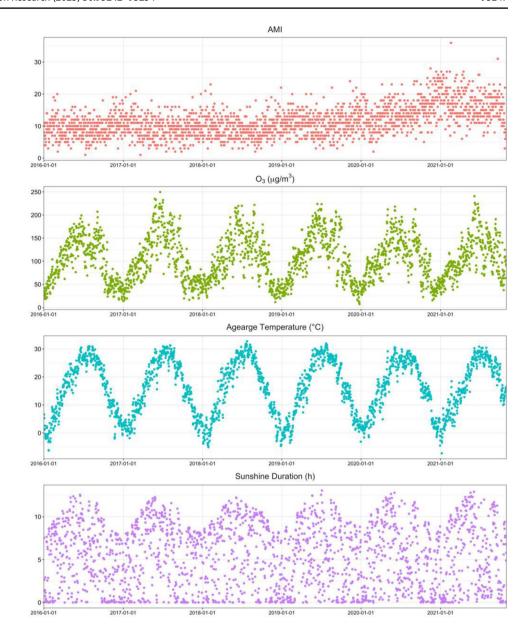
### Discussion

This study was performed in the north-central provinces of China, with four seasons of the year. It is representative of China as a major agricultural province with low per capita income levels, and is one of the most polluted provinces in China (Zhang et al. 2022). We first constructed single-pollution models for different air pollutants of  $O_3$ ,  $PM_{2.5}$ ,

PM<sub>10</sub>, CO, NO<sub>2</sub>, and SO<sub>2</sub> and the risk of AMI hospitalization (Table S4), and found that short-term  $O_3$  exposure was associated with a more significant increase in the risk of AMI hospitalization than other air pollutants, and therefore, the analysis was focused on short-term O<sub>3</sub> pollution exposure. We found that AMI hospitalization and shortterm O<sub>3</sub> exposure in Henan were significantly associated with a lag effect, with the greatest risk at a cumulative lag of 3 days (RR = 1.132, 95% CI:1.083-1.182) and a linear exposure–response curve (Figure S3). In addition, males, younger patients, warm season, and LSD groups were more susceptible to O<sub>3</sub> levels. To our knowledge, this is the first time that DLMN analysis has been used to explore the relationship between short-term O<sub>3</sub> exposure and AMI hospitalization in China and to attempt to quantify the number of O<sub>3</sub>-attributable AMI hospitalizations, hospitalization days, and hospitalization costs.



Fig. 1 The distribution of  $O_3$ , meteorological variables, and AMI hospitalizations in Henan, China, during 2016–2021. AMI, acute myocardial infarction;  $O_3$ , ozone



In epidemiological studies,  $O_3$  exposure is associated with the risk of cardiovascular and cerebrovascular diseases and mortality (Nuvolone et al. 2013; Raza et al. 2019; Wen et al. 2022; Zhang et al. 2019, 2021). A multicenter study from Italy showed that short-term exposure to  $O_3$  increased all-cause mortality in a population with an RR of 1.0043 (95% CI:1.0029, 1.0057) (Wen et al. 2022). Raza et al. (2019) found that short-term exposure to  $O_3$  increases the risk of out-of-hospital cardiac arrest. Henrotin et al. (2010) showed that short-term exposure to even low levels of  $O_3$  may cause ischemic cerebrovascular events and myocardial infarction (MI) and was most significant at a cumulative lag of 3 days with an RR of 1.115 (95% CI:1.027–1.209), consistent with our findings; however, Henrotin et al. included fewer patients

with MI. Furthermore, it has been shown that there is no significant association between hourly  $O_3$  exposure and the onset of acute coronary syndromes (Chen et al. 2022), which may be due to various reasons, such as the study period, study site, and differences in  $O_3$  exposure levels. Most previous studies on  $O_3$  and cardiovascular disease were conducted in developed countries, where the mean 8-h  $O_3$  concentrations were generally below 60  $\mu$ g/m³, which is significantly lower than the mean 8-h  $O_3$  concentration in our study area (104.3  $\mu$ g/m³). The outcome supports this study's results and points to a need for further control of  $O_3$  levels to lower levels in our study area.

Because pollutants in the atmosphere usually exist simultaneously, we constructed a dual-pollution model for



Table 2 RR (95% CI) of AMI hospitalizations expose to O<sub>3</sub> single- and dual-pollution model (every 10 μg/m<sup>3</sup> increase) at different lag days

	$O_3$		$O_3 + PM_{2.5}$	2.5	$O_3 + PM_{10}$	01	$O_3 + CO$		$O_3 + NO_2$		$O_3 + SO_2$	
	RR	95% CI	RR	95% CI	RR	95% CI	RR	95% CI	RR	95% CI	RR	95% CI
Lag0	1.076	(1.043–1.111)	1.074	(1.035–1.114)	1.069	(1.031–1.109)	1.077	(1.038–1.117)	1.062	(1.024–1.102)	1.060	(1.022–1.099)
Lag1	1.037	(1.022-1.053)	1.033	(1.015-1.051)	1.031	(1.013-1.049)	1.032	(1.014-1.05)	1.030	(1.012-1.048)	1.034	(1.016-1.051)
Lag2	1.013	(0.997-1.029)	1.008	(0.989-1.027)	1.008	(0.989-1.027)	1.006	(0.987-1.025)	1.010	(0.991-1.029)	1.015	(0.996-1.034)
Lag3	1.000	(0.988-1.013)	0.995	(0.980-1.010)	966.0	(0.981-1.011)	0.993	(0.979-1.008)	0.999	(0.984-1.014)	1.003	(0.988-1.018)
Lag4	0.995	(0.983-1.008)	0.99	(0.975-1.004)	0.991	(0.977-1.006)	0.989	(0.974-1.003)	0.993	(0.979-1.008)	0.995	(0.981-1.01)
Lag5	0.995	(0.980-1.011)	0.989	(0.971-1.007)	0.660	(0.972-1.009)	0.988	(0.97-1.006)	0.991	(0.973-1.009)	0.991	(0.973-1.009)
Lag6	0.997	(0.985-1.01)	0.988	(0.973-1.003)	0.990	(0.975-1.005)	0.986	(0.971-1.001)	0.988	(0.973-1.003)	0.989	(0.974-1.004)
Lag7	0.999	(0.973-1.025)	0.984	(0.955-1.015)	986.0	(0.957-1.017)	0.978	(0.949-1.009)	0.983	(0.953-1.013)	0.987	(0.958-1.018)
Lag01	1.116	(1.073-1.161)	1.109	(1.060-1.160)	1.102	(1.054-1.153)	1.111	(1.062-1.162)	1.094	(1.046-1.144)	1.095	(1.047-1.146)
Lag02	1.131	(1.085-1.179)	1.118	(1.066-1.173)	1.111	(1.059-1.165)	1.117	(1.065-1.172)	1.104	(1.053-1.158)	1.112	(1.060-1.166)
Lag03	1.132	(1.083-1.182)	1.112	(1.058-1.169)	1.107	(1.053-1.163)	1.110	(1.056-1.167)	1.103	(1.049-1.160)	1.115	(1.061-1.172)
Lag04	1.126	(1.077-1.178)	1.101	(1.046-1.158)	1.097	(1.042-1.154)	1.097	(1.043-1.154)	1.096	(1.041-1.153)	1.110	(1.054-1.168)
Lag05	1.121	(1.070-1.175)	1.088	(1.031-1.148)	1.086	(1.029-1.146)	1.084	(1.027-1.143)	1.086	(1.029-1.146)	1.099	(1.042-1.161)
Lag06	1.118	(1.064-1.176)	1.075	(1.016-1.137)	1.075	(1.016-1.138)	1.068	(1.009-1.130)	1.073	(1.014-1.136)	1.087	(1.027-1.151)
Lag07	1.117	(1.064–1.172)	1.058	(1.002–1.117)	1.061	(1.004–1.120)	1.045	(0.989–1.103)	1.054	(0.998–1.114)	1.073	(1.016–1.134)

Bold: p < 0.05.



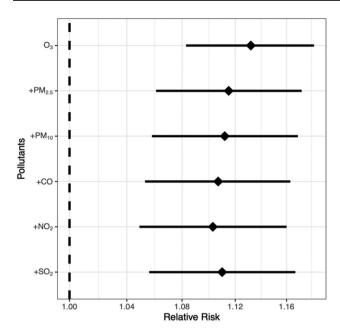


Fig. 2 RR (95% CI) of AMI hospitalizations for every 10  $\mu$ g/m.<sup>3</sup> increase in O<sub>3</sub> in the dual-pollutant model at lag 03

PM<sub>2.5</sub>, PM<sub>10</sub>, CO, NO<sub>2</sub> and SO<sub>2</sub>, and included covariates such as temperature, humidity, Dow, and public holidays in the model. The results showed no significant difference compared with the single-pollution model for O<sub>3</sub>, and the RR of the dual-pollution model was slightly lower than that of the single-pollution model, probably due to the significant negative correlation between O<sub>3</sub> and other air pollutants (Table S3). O<sub>3</sub> and temperature were significantly positively correlated (r=0.78), with significantly higher O<sub>3</sub> concentrations in the warm season than in the cold season (Table S6). The results of the subgroup analysis indicated that short-term exposure to O<sub>3</sub> was associated with a significantly higher risk of AMI hospitalization in the warm season than in the cold season, similar to the findings of Wen et al. (2022). Similarly, we found that  $O_3$  and sunshine duration were also positively correlated (r = 0.54), which may be related to the involvement of ultraviolet light in the formation of O<sub>3</sub> (Chang et al. 2022; Wang et al. 2020b). As we found that the association between O<sub>3</sub> and the risk of AMI hospitalization was insignificant in the cold season, we further divided the patients into the LSD and SSD groups according to the mean sunshine hours in the warm season (6.33 h) (Table S7). We found that patients in the LSD group were associated with a higher risk of AMI, supporting the association between O<sub>3</sub> concentration and the risk of AMI hospitalization. Concerning the age and sex of the susceptible population, the findings of current studies are mostly inconsistent; Wen et al. in Italy did not find differences between age and sex groups exposed to O<sub>3</sub> and all-cause mortality (Wen et al. 2022). In a multi-community study,

Schwartz et al. reported that for every 10 µg/m<sup>3</sup> increase in O<sub>3</sub> (mean 0–2 day lag), cardiovascular disease mortality was higher in people aged ≥ 65 years than in younger people, and females and blacks were more likely to be affected (Medina-Ramón and Schwartz 2008). In this study, we have found that males and younger patients are more likely to be affected by  $O_3$ , possibly because there were more males and patients with AMI aged 14-64 years in this study and that this population would be more likely to expose themselves to higher levels of O<sub>3</sub> through outdoor activities or by engaging in occupations (such as policeman and delivery-man) (Chang et al. 2022); at the meantime, males were more likely to suffer from respiratory disease (Wang et al. 2022b), which in turn affects cardiorespiratory function, which may also increase the risk of AMI hospitalization due to O<sub>3</sub> exposure in this population.

The mechanism of O<sub>3</sub> and the occurrence of AMI is unclear, and the results of some studies are inconsistent (Goodman et al. 2015). Michaudel et al. found that  $O_3$  dissolved in plasma generates reactive oxygen species (Michaudel et al. 2018), causing apoptosis of endothelial cells and damaging blood vessels (Emon et al. 2021). In addition, a study on healthy individuals showed that exposure to O<sub>3</sub> resulted in elevated plasma levels of inflammatory markers (Kim et al. 2011; Pirozzi et al. 2015) and that inflammatory mediators can enter the circulatory system and stimulate platelet activation (Day et al. 2017) and the release of coagulation factors (Münzel et al. 2017), increasing the chances of thrombosis and promoting coronary events. O<sub>3</sub> exposure may also be associated with changes in the autonomic nervous system (Arjomandi et al. 2015; Jia et al. 2011) and blood pressure (Niu et al. 2022); however, some studies have found a protective effect of O<sub>3</sub> on ischemic myocardium (Wang et al. 2022a; Yu et al. 2021).

To complement the traditional exposure risk assessment, we evaluated the number of AMI hospitalizations, hospitalization days, and hospitalization costs attributable to  $O_3$ . The results showed that 11.66% of AMI hospitalizations were attributable to the effects of O<sub>3</sub> in the environment, equivalent to 2856 patients and 90 million RMB (Table 3). A modelling study of 12,946 cities included globally showed that the fraction of global O<sub>3</sub>-attributable mortality occurring in urban areas gained in their share of global O<sub>3</sub>-attributable burden from 2000 to 2019 (from 35 to 37%) (Malashock et al. 2022). Another study of the global burden of chronic obstructive pulmonary disease (COPD) attributable to ambient O<sub>3</sub> exposure, conducted in 204 countries around the world, showed that among COPD-related deaths, attributable O<sub>3</sub> exposure is generally on the rise globally, from 8.22% in 1990 to 11.13% in 2019 (Wang et al. 2022b). In contrast, a study by Chen et al. (2017b) in area of Zhejiang Province, China, found that the population attributable fractions for  $O_3$  exposure to COPD is 5.32%. This study fills a research



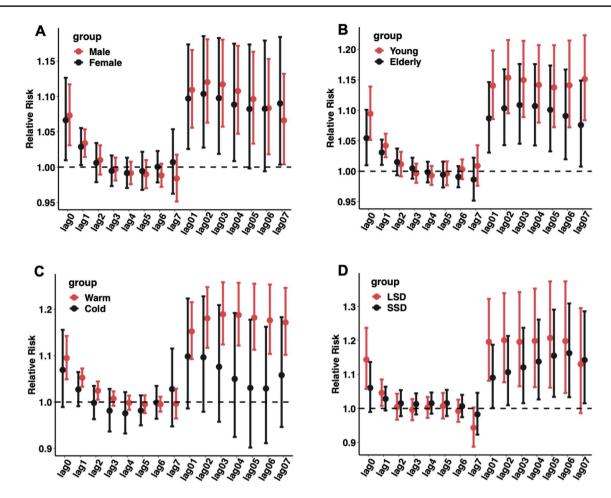


Fig. 3 RR (95% CI) of AMI hospitalizations expose to  $O_3$  (every  $10 \mu g/m^3$  increase) in different subgroups at different lag days. A The single-day and cumulative relative risk effects in males and females. B The single-day and cumulative relative risk effects in young and

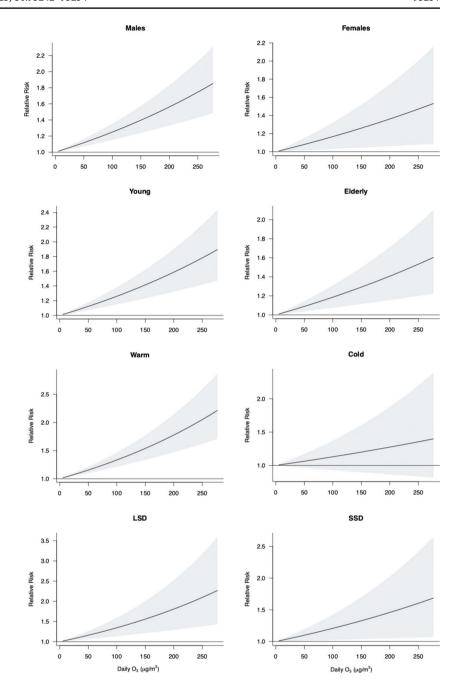
elderly.  $\bf C$  The single-day and cumulative relative risk effects in warm season and cold season.  $\bf D$  The single-day and cumulative relative risk effects of LSD (long sunshine duration) and SSD (short sunshine duration) in warm season

gap in China in terms of AMI hospitalizations attributable to O<sub>3</sub> exposure. The medical burden on local hospitals is enormous, and the local economic loss to the government and patients is huge. The Chinese government promulgated the Ambient Air Quality Standard (GB 3095–2012) in 2012, in which the secondary standard applicable to general areas sets the 8-h average concentration of O<sub>3</sub> at 160 µg/m<sup>3</sup>; even below this standard, short-term exposure to O<sub>3</sub> exhibits a significant association with an increased risk of AMI hospitalization. In September 2021, the World Health Organization released the latest unprecedentedly stringent "Air Quality Guidelines" (AQG) (Burki 2021), which set higher requirements for atmospheric O<sub>3</sub> concentrations and added a new indicator of O<sub>3</sub> concentrations less than 60 μg/m<sup>3</sup> in the warm season, compared to the warm season in Henan, where O<sub>3</sub> concentrations are much higher than this standard. If more stringent measures could be implemented to control O<sub>3</sub> to lower concentrations, as shown in Figure S4, hundreds of AMI cases could be prevented yearly, and millions of RMB could be saved. This study is the first to analyze the relationship between short-term exposure to  $O_3$  and the risk of AMI hospitalization in areas of China with more severe  $O_3$  pollution, providing a basis for the prevention of AMI and environmental improvement in Henan Province.

Our study had some limitations. First, as an environmental ecological study, we could not include more comprehensive information on individuals in our study, such as disease history, complications, biochemical indicators, and medication use, which will need to be included in future in-depth clinical studies. Second, owing to the nature of the timeseries study design and the uneven distribution of air monitoring stations in some study areas, there was variability in O<sub>3</sub> exposure per patient, which may have led to reduced variability in exposure conditions and biased results. Third, in practical studies, the AF calculations we performed were often not univariate, and also, observational studies introduce certain confounding effects. This leads to some deviation from the true AF when we calculate AF. Fourth,



Fig. 4 Exposure—response curves in different subgroups of  $O_3$  with AMI hospitalizations at lag 03. LSD, long sunshine duration (sunshine duration > 6 h); SSD, short sunshine duration (sunshine duration  $\leq$  6 h); both LSD and SSD are warm season sunshine durations



 $\begin{tabular}{ll} \textbf{Table 3} & Fraction and number of \\ AMI hospitalizations, hospital \\ stays, and hospitalization costs \\ attributable to $O_3$ \\ \end{tabular}$ 

	Attributable fraction of hospitalizations (%)		Attributable number of hospitalizations (no.)		Attributable number of hospital stays (days)		Attribut- able number of hospitalization costs (RMB, million)	
Total	11.66	(7.66–15.40)	2856	(1878–3768)	33,492	(22,014–44,226)	90	(60–120)
Males	10.47	(5.39–15.25)	702	(360–1020)	8,340	(4296–12,144)	18	(12–30)
Females	8.93	(1.86-15.47)	1590	(330-2754)	18,528	(3879-32,112)	54	(12-90)
Young	13.04	(8.17-17.63)	1476	(924-1992)	17,418	(10,914–23,538)	42	(24-60)
Elderly	9.83	(4.31–14.97)	1296	(570-1968)	15,102	(6618-22,992)	42	(18-66)
Warm	15.97	(11.03-20.57)	1620	(1122–2088)	18,978	(13,110–24,450)	48	(36–66)



we did not consider information related to hospitalization reimbursement for each patient with AMI when calculating the annual hospitalization costs that could be reduced by reducing O<sub>3</sub> and the fact that reimbursement may vary across regions, which may have affected our estimates of O<sub>3</sub>-induced costs. Fifth, this study was conducted in only one province in central China; therefore, caution should be taken when generalizing to populations in other regions or countries.

# **Conclusions**

We found that short-term exposure to  $O_3$  was significantly associated with an increased risk of hospitalization for AMI and that the exposure–response curve was linear at a cumulative lag of 3 days, particularly for males, younger patients, warm season, and long sunshine duration with greater risk. Further reductions in  $O_3$  levels, particularly in developing countries, may bring substantial health and economic benefits to patients and local healthcare institutions.

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Author contribution Yulong Wei: conceptualization, methodology, data curation, formal analysis, writing — original draft; Lin Fei: funding acquisition, writing — review and editing; Yongbin Wang: conceptualization, methodology; Min Zhang: conceptualization, methodology, writing — review and editing; Zhigang Chen: conceptualization, methodology; Huige Ge: conceptualization, methodology, data curation; Shiqi Ge: conceptualization, methodology, data curation; Sen Zhu: conceptualization, methodology; Pingshuan Dong: conceptualization; Kan Yang: conceptualization; Na Xie: conceptualization; Guoan Zhao: conceptualization, funding acquisition, writing — review and editing.

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**Data availability** All data were presented in our analytical results or please contact the first author or the corresponding author on reasonable request.

#### **Declarations**

Consent to participate Not applicable.

**Consent for publication** The manuscript was approved by all authors.

Competing interests The authors declare no competing interests.

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