

THE LAGGED EFFECT OF AIR POLLUTION ON HUMAN EOSINOPHILS: A DISTRIBUTED LAG NON-LINEAR MODEL

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Abstract

Objectives: The aim of this study was to determine the lag between exposure to air pollutants and changes in human eosinophil counts. **Material and Methods:** This was a retrospective study employing 246 425 physical examination records dated December 2013 – December 2016 from Chengdu, China. The authors determined the prevalence of individuals with eosinophil counts above the normal reference range each day. A distributed lag non-linear model was used to evaluate the lagged effect of each air pollutant on eosinophil counts. The lagged effects of each air pollutant were counted and presented with smoothing splines. **Results:** The effects of air pollutants such as particulate matter (PM_{2.5}, aerodynamic diameters <2.5 μm; PM₁₀, aerodynamic diameters <10 μm), nitrogen dioxide (NO₂) and ozone (O₃) were evaluated. In women, the effects of PM_{2.5} (RR = 1.154, 95% CI: 1.061–1.255) and PM₁₀ (RR = 1.309, 95% CI: 1.130–1.517) reached the maximum values on lag day 0. In men, there was no significant effect of PM_{2.5}, but significant effects of PM₁₀ were found for lag days 20–28. The effects of NO₂ and O₃ on eosinophils were not statistically significant for either gender. **Conclusions:** The air pollutants of PM₁₀ have a significant effect on human eosinophils for both women and men, but with different temporal patterns, with women showing a lag of 0–5 days and men showing a lag of 20–28 days. In addition, PM_{2.5} was significant for women with a lag of 0–3 days but it was not significant for men. *Int J Occup Med Environ Health.* 2020;33(3):299–310

Key words:

air pollution, gender difference, eosinophils, real-world data, lagged effect, distributed lag non-linear model

INTRODUCTION

According to the World Health Organization (WHO) report of 2018, 9 out of 10 people now breathe polluted air, which kills 7 million people every year [1]. With the rapid development of industrialization, air pollution has be-

come one of the most important environmental problems in China. The Chinese Department of Environmental Protection considered particulate matter (PM_{2.5}, aerodynamic diameters <2.5 μm; PM₁₀, aerodynamic diameters <10 μm), sulfur dioxide (SO₂), nitrogen dioxide (NO₂),

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carbon monoxide (CO) and ozone (O₃) as the primary air pollutants and included them in the latest China's Ambient Air Quality Standard (GB 3095-2012) in 2012 [2]. The 24-h mean threshold values of PM_{2.5}, PM₁₀, SO₂, CO and O₃ in GB 3095-2012 are higher than the 2005 WHO air quality suggested values.

Previous studies have confirmed that exposure to air pollutants would lead to the development of cardiovascular diseases, lung diseases, and immunity-related diseases such as allergic rhinitis, childhood asthma, and atopic dermatitis [3–8]. Epstein et al. [9] evaluated the relationship between exposure to traffic pollutants and asthma control, and found that traffic pollutants increased eosinophilic inflammation in older adults with poorly controlled asthma. Meanwhile, other authors reported that eosinophils increased with exposure to air pollutants [10–12]. On the other hand, Fauzie and Venkataramana [13] performed animal experiments with albino mice exposed to traffic-related air pollutants, and the results showed that eosinophils were significantly reduced.

However, some controversy remains about the impact of air pollution on eosinophils. One study of the association between eosinophilic activation and air pollution showed no significant associations between the number of eosinophils and PM_{2.5} or NO₂, which did not show the cumulative effect of air pollution [14].

Stimulated by the above studies, physical examination records, and air pollutant and meteorological data are used in this study to determine the effect of air pollution on human eosinophil counts and clarify whether there is a lagged effect of air pollution on human eosinophils.

MATERIAL AND METHODS

Study area

Chengdu is located in the western part of the Sichuan Province, geographically situated between 102°54' E ~ 104°53' E and 30°05' N ~ 31°26' N. In 2016, the city's land area was 14 335 km², and there were 11 urban built-up ar-

reas totaling about 4000 km². Due to high traffic density and special basin topography, the urban area suffers from a high concentration of air pollutants [15,16].

Study population

This is a population-based study with a retrospective analysis of real-world data for the period of December 2013 – December 2016. The data were extracted from the records of men and women sent by their employers to receive annual physical examinations at the Physical Examination Center of the Sichuan Provincial People's Hospital. The biochemical tests of all subjects were carried out in the Department of Clinical Laboratory in the Sichuan Provincial People's Hospital immediately after sample collection. The normal reference range for serum eosinophil counts, as provided by the clinical laboratory, is 0.02–0.52 × 10⁹/l.

Initially 275 088 records were included. In order to eliminate abnormal interference, the following exclusion criteria were used:

- incomplete data, for example, the examination time, specific age and gender, and examination results are missing;
- individuals aged <18 or ≥65 years;
- eosinophils <0.02 × 10⁹/l;
- diagnosis of cancer;
- dates with <15 physical examination records.

Following these exclusions, 246 425 records remained. The validity and reliability of the physical examination data were ensured through a stringent quality assurance and quality control program by the Sichuan Provincial People's Hospital. The Ethics Committee in the Sichuan Provincial People's Hospital gave its permission for this study.

Air pollutants and meteorological data

All air pollutant and meteorological data were received from the China National Environmental Monitoring Center [17]. The meteorological data, including

the mean value of daily temperature (DT), relative humidity (RH), daily average values of $PM_{2.5}$, PM_{10} , SO_2 , NO_2 , CO concentration, and 8-h maximum daily sliding average of O_3 concentration in Chengdu, were all recorded throughout the study period. The methods for air pollutant and meteorological data collection, including time resolution and instrumentation details, were as described previously [16].

Statistical analysis

The authors assumed that the impact of air pollutants on the human body was not significantly effective beneath a certain threshold. That threshold was selected in accordance with China's Ambient Air Pollution Standards (GB 3095-2012) [2]. A distributed lag non-linear model (DLNM) was used to evaluate the lagged effects of each air pollutant [18]. The authors counted the number of physical examination records per day with eosinophil counts exceeding the upper limit of the normal reference range ($0.52 \times 10^9/l$) as the daily prevalence.

Due to the observed dispersion of data, the authors used the quasi-Poisson regression to fit them. The DLNM equation was as follows:

$$\log(E(Y_t)) = \alpha + cb(X_t, l, thr) + ns(time, df) + ns(DT, df) + ns(RH, df) + DOW \quad (1)$$

where:

Y_t – the event rates of the parts per 10 000 by every day in the study period;

$E((Y_t))$ – the expected value for Y_t ;

α – the intercept;

cb – the cross-basis function which can be depicted as a 2-dimensional space of functions describing simultaneously the shape of the relationship along a single air pollutant and its distributed lag effects [19];

X_t – the daily average concentration of each air pollutant ($PM_{2.5}$, PM_{10} , SO_2 , NO_2 , CO and O_3), which depicted a linear effect

on human eosinophils (the lag effect of X_t was depicted using a third-degree polynomial function);

l – the number of lag days for the effect of each air pollutant;

thr – the threshold of each air pollutant effect;

ns – the natural spline function which was used to smoothen the time trend effects of a long-term character, daily temperature and relative humidity;

$time$ was used to control the time trend effects of a long-term character;

df – the degree of freedom;

DT – daily temperature;

RH – relative humidity;

DOW – the day of the week which was used to control the week effects.

The authors chose the appropriate df according to the Partial Autocorrelation Function (PACF), and the calculation method was as reported previously [20]. The maximum permitted lag was 35 days, which was determined by the survival time of eosinophils and a smaller value of the quasi-Akaike information criterion (QAIC) [21]. Relative risk (RR) values with their respective 95% confidence intervals (95% CI) were used to show the effect of each air pollutant with a $10 \mu g/m^3$ increase in concentration.

Data processing and statistical analyses were performed using “splines” and “dlm” packages of R software (version 3.5.2) [18,22].

RESULTS

Basic data description

Table 1 shows the basic information derived from physical examination records. After excluding extreme values and filtering data according to the exclusion criteria, 246 425 physical examination records were included in this study. Among these, 4546 records showed eosinophil counts exceeding $0.52 \times 10^9/l$, of which 3164 concerned men and 1382 women. The prevalence in men was significantly

Table 1. Characteristics of the medical records of 246 425 participants by gender at the Physical Examination Center of the Sichuan Provincial People's Hospital in Chengdu, China (December 2, 2013 – December 2, 2016)

Variable	Participants (N = 246 425)		
	total	men (N = 136 030)	women (N = 110 395)
Records			
physical examination [n]	246 425	136 030	110 395
daily (Me (25–75%))	266 (199–348)	145 (104–193)	120 (89–151)
Age [years] (M±SD)	40.3±11.20	40.7±11.11	39.8±11.42
BMI [kg/m ²] (M±SD)	23.3±3.40	24.3±3.2	22.0±3.10
Blood cell counts			
erythrocytes [10 ¹² /l] (M±SD)	4.86±0.55	5.16±0.48	4.49±0.40
leukocytes [10 ⁹ /l] (Me (25–75%))	6.02 (5.012–7.09)	6.27 (5.37–7.35)	5.72(4.86–6.73)
neutrophils [10 ⁹ /l] (Me (25–75%))	3.40 (2.77–4.17)	3.50 (2.88–4.27)	3.27 (2.64–4.04)
eosinophils [10 ⁹ /l] (Me (25–75%))	0.11 (0.07–0.18)	0.13 (0.08–0.21)	0.09 (0.06–0.15)
basophils [10 ⁹ /l] (Me (25–75%))	0.032 (0.021–0.048)	0.037 (0.022–0.005)	0.03 (0.02–0.042)
lymphocytes [10 ⁹ /l] (Me (25–75%))	1.95 (1.60–2.37)	2.04 (1.67–2.47)	1.85 (1.53–2.23)
monocytes [10 ⁹ /l] (M±SD)	0.43±0.14	0.46±0.15	0.38±0.12

higher than in women according to the Wilcoxon signed-rank test (0.13 vs. 0.09, $p < 0.01$).

Variable description

The daily average values of PM_{2.5}, PM₁₀, SO₂, NO₂, CO concentration, the 8-h maximum daily sliding average of O₃ concentration, and the daily prevalence of eosinophil counts exceeding the upper limit of the reference range are listed in Table 2. The daily average concentrations of PM_{2.5} and PM₁₀ for 330 and 234 days, respectively, were higher than the standard values stipulated in China's Ambient Air Pollution Standards. The values of NO₂ and O₃ exceeded the standard values for 64 and 132 days, respectively. However, the values of SO₂ and CO were always under the standard values in the study period, and thus DLNM could not evaluate their effects. Meteorological data, including temperature and relative humidity, were dispersed evenly.

Air pollution effects

The effects of 4 pollutants (PM_{2.5}, PM₁₀, NO₂ and O₃) on men and women are shown in Figure 1. Of these pollutants, PM_{2.5} and PM₁₀ had a greater effect on eosinophils in men than in women, and showed a distinct lagged effect in men (Figure 1a and Figure 1b vs. Figure 1e and Figure 1f).

In women, the effects of PM_{2.5} (RR: 1.154, 95% CI: 1.061–1.255) and PM₁₀ (RR: 1.309, 95% CI: 1.130–1.517) reached the maximum value on lag day 0, then decreased as the lag time increased (Figure 1e and Figure 1f). More specifically, PM_{2.5} and PM₁₀ produced effects in women on lag days 0–3 and lag days 0–5, respectively. In men, the results for PM_{2.5} were not significant. However, significant results were found for PM₁₀ on lag days 20–28 (RR: 1.072, 95% CI: 1.005–1.143), with the maximum value occurring on lag day 24.

The other air pollutants, NO₂ and O₃, had no significant effect on eosinophil counts in men (Figure 1c and Figure 1d)

Table 2. Summary statistics of daily prevalence, air pollutants and weather factors at the Physical Examination Center of the Sichuan Provincial People's Hospital in Chengdu, China (December 2, 2013 – December 2, 2016)

Exposure variable	Me (25–75%)	Min.	Max
Daily incidence rate			
total	176 (118–245.2)	0	938
men	220 (132–321)	0	1250
women	109 (0–190)	0	1667
Daily pollutant concentration			
PM _{2.5} [$\mu\text{g}/\text{m}^3$] ^a	53 (36–83)	10	396
PM ₁₀ [$\mu\text{g}/\text{m}^3$] ^b	91 (63–140)	16	562
SO ₂ [$\mu\text{g}/\text{m}^3$] ^c	14 (11–20)	4	61
NO ₂ [$\mu\text{g}/\text{m}^3$] ^d	50 (41–62)	15	118
O ₃ [$\mu\text{g}/\text{m}^3$] ^e	82 (50–127)	7	293
CO [mg/m^3] ^f	1 (0.9–1.3)	0.4	2.6
Weather factor			
temperature [$^{\circ}\text{C}$]	20 (12–24)	2	32
relative humidity [%]	73 (66–82)	27	100

^a Pollution days: 330 above the standard value, 767 under the standard value.

^b Pollution days: 234 above the standard value, 863 under the standard value.

^c Pollution days: 0 above the standard value, 1097 under the standard value.

^d Pollution days: 64 above the standard value, 1033 under the standard value.

^e Pollution days: 132 above the standard value, 965 under the standard value.

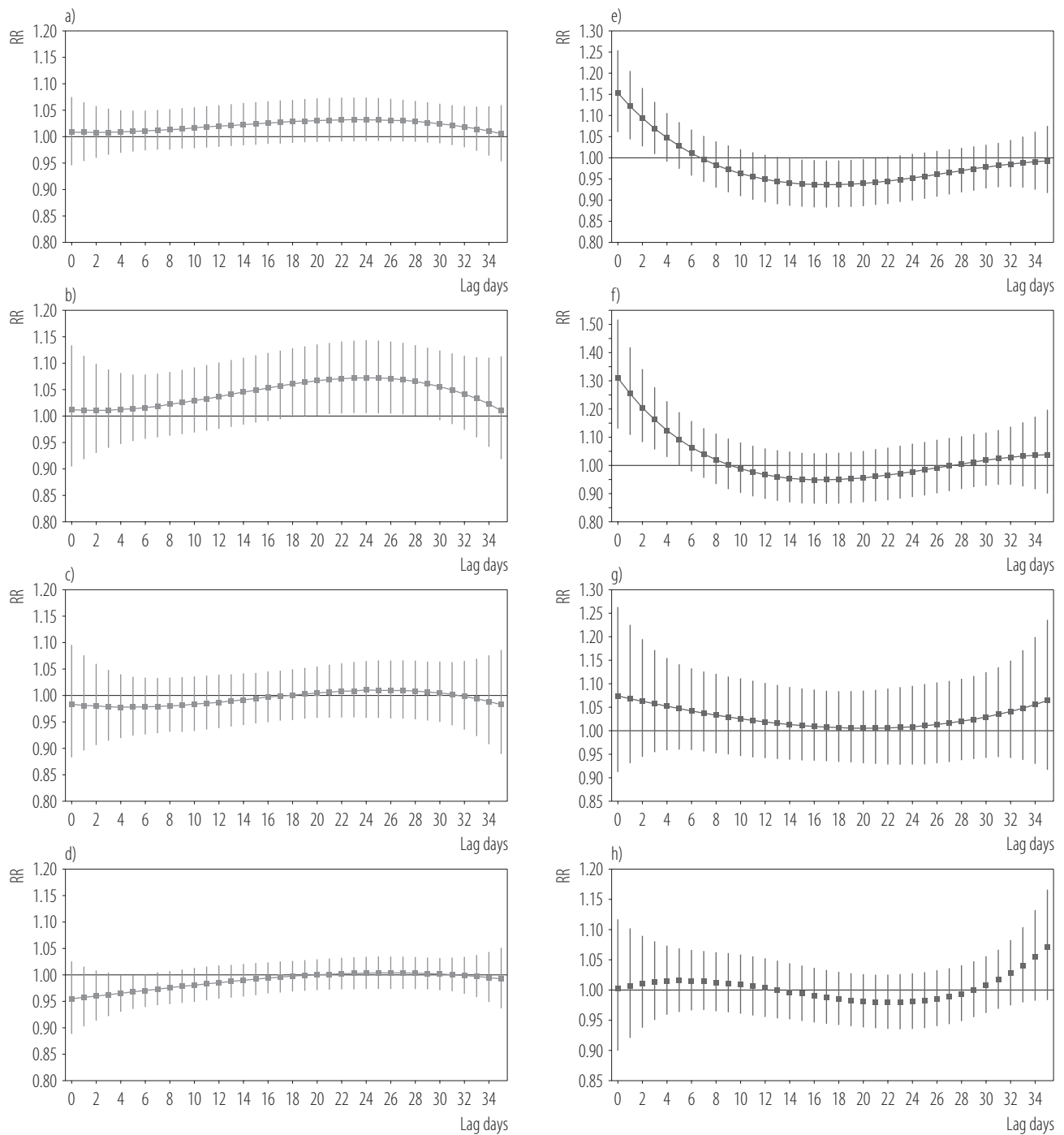
^f Pollution days: 0 above the standard value, 1097 under the standard value.

or women (Figure 1g and Figure 1h). More data can be found in Table 3.

DISCUSSION

Clinically, human eosinophils can be used to indicate the degree of immune response to a disease [23,24]. Eosinophil counts are especially good indicators for hypersensitivity pneumonitis, eczema, and helminth infections. Here, the authors used eosinophil counts as a biomarker to show the harmful effects of air pollution. The effect of exposure to air pollution on the human body is usually a lagged effect that is not limited to the period when it is observed [17,18]. Therefore, DLNMs were used to demonstrate the effects of 4 pollutants (PM_{2.5}, PM₁₀, NO₂ and O₃) on eosinophil counts. To date, DLNMs have been widely used to discover lagged effects of exposure (especially for air pollution) on outcomes [25–27].

For both men and women, PM_{2.5} and PM₁₀ had the largest effects on eosinophil counts. In 2010, Sehlstedt et al. [12] took 15 healthy subjects, exposed them to an average particulate matter concentration of 270 $\mu\text{g}/\text{m}^3$ or filtered air for 1 h, and found that particulate matter exposure increased the number of bronchoalveolar lavage eosinophils in these healthy subjects. Moreover, Carlsten et al. [10] recruited 18 blinded atopic volunteers and had them inhale 300 $\text{mg PM}_{2.5}/\text{m}^3$ of diesel exhaust; the results showed that airway eosinophils increased in atopic patients at environmentally relevant concentrations of PM_{2.5}. In the 2 studies mentioned above, the effects of air pollution on gender were not evaluated, and neither were lagged effects. A separate study reported no significant association between the number of eosinophils and exposure to PM_{2.5} in 521 children [14]. However, this study used adults aged



Vertical bars – 95% CI.

NO₂ – nitrogen dioxide; O₃ – ozone; PM_{2.5} – particulate matter with an aerodynamic diameter of <2.5 μm; PM₁₀ – particulate matter with an aerodynamic diameter of <10 μm.

Figure 1. The lagged effect of air pollution on eosinophils in a–d) men and e–h) women – splines showed changes in relative risk (RR) values caused by a 10 μg/m³ increase in the average daily concentration of a) and e) PM_{2.5}, b) and f) PM₁₀, c) and g) NO₂, and the 8-h max daily sliding average of d) and h) O₃ during 35 lag days

Table 3. Air pollution effect on men and women at the Physical Examination Center of the Sichuan Provincial People's Hospital in Chengdu, China (December 2, 2013– December 2, 2016)

Lag days	Air pollutant							
	PM _{2.5} [$\mu\text{g}/\text{m}^3$]	95% CI	PM ₁₀ [$\mu\text{g}/\text{m}^3$]	95% CI	NO ₂ [$\mu\text{g}/\text{m}^3$]	95% CI	O ₃ [$\mu\text{g}/\text{m}^3$]	95% CI
Men								
lag0	1.00856	0.94598–1.07527	1.01202	0.90334–1.13378	0.98292	0.88146–1.09605	0.95417	0.88853–1.02466
lag1	1.00833	0.95412–1.06562	1.01075	0.91741–1.11358	0.98069	0.89443–1.07526	0.95677	0.90180–1.01510
lag2	1.00841	0.96076–1.05841	1.01029	0.92928–1.09836	0.97906	0.90503–1.05915	0.95940	0.91311–1.00804
lag3	1.00875	0.96592–1.05348	1.01059	0.93890–1.08775	0.97800	0.91323–1.04738	0.96205	0.92237–1.00344
lag4	1.00934	0.96971–1.05059	1.01157	0.94633–1.08130	0.97746	0.91915–1.03947	0.96471	0.92960–1.00115
lag5	1.01015	0.97236–1.04940	1.01316	0.95188–1.07839	0.97739	0.92316–1.03480	0.96736	0.93497–1.00087
lag6	1.01115	0.97419–1.04951	1.01531	0.95603–1.07827	0.97775	0.92578–1.03263	0.97000	0.93889–1.00214
lag7	1.01232	0.97550–1.05052	1.01795	0.95931–1.08017	0.97849	0.92761–1.03216	0.97262	0.94186–1.00439
lag8	1.01363	0.97657–1.05209	1.02100	0.96221–1.08339	0.97958	0.92915–1.03275	0.97520	0.94432–1.00709
lag9	1.01506	0.97758–1.05397	1.02442	0.96509–1.08740	0.98098	0.93077–1.03389	0.97774	0.94662–1.00988
lag10	1.01657	0.97864–1.05597	1.02813	0.96814–1.09183	0.98263	0.93267–1.03528	0.98022	0.94894–1.01253
lag11	1.01815	0.97981–1.05800	1.03206	0.97147–1.09644	0.98451	0.93491–1.03674	0.98264	0.95138–1.01493
lag12	1.01977	0.98107–1.05998	1.03615	0.97505–1.10109	0.98656	0.93749–1.03820	0.98498	0.95394–1.01702
lag13	1.02139	0.98242–1.06191	1.04034	0.97882–1.10572	0.98875	0.94032–1.03968	0.98723	0.95659–1.01885
lag14	1.02300	0.98381–1.06375	1.04454	0.98268–1.11029	0.99103	0.94326–1.04121	0.98938	0.95925–1.02045
lag15	1.02457	0.98520–1.06552	1.04870	0.98651–1.11481	0.99336	0.94618–1.04289	0.99142	0.96184–1.02191
lag16	1.02607	0.98652–1.06720	1.05273	0.99018–1.11924	0.99569	0.94892–1.04477	0.99334	0.96426–1.02330
lag17	1.02747	0.98774–1.06879	1.05658	0.99357–1.12358	0.99799	0.95136–1.04691	0.99513	0.96642–1.02469
lag18	1.02874	0.98882–1.07027	1.06015	0.99659–1.12776	1.00021	0.95338–1.04933	0.99677	0.96825–1.02612
lag19	1.02987	0.98975–1.07161	1.06339	0.99919–1.13172	1.00229	0.95495–1.05198	0.99825	0.96973–1.02762
lag20	1.03081	0.99050–1.07276	1.06621	1.00131–1.13532	1.00421	0.95607–1.05478	0.99957	0.97084–1.02916
lag21	1.03155	0.99109–1.07366	1.06854	1.00294–1.13843	1.00591	0.95676–1.05759	1.00072	0.97162–1.03068
lag22	1.03205	0.99152–1.07424	1.07031	1.00409–1.14089	1.00735	0.95711–1.06023	1.00167	0.97215–1.03208
lag23	1.03229	0.99179–1.07444	1.07143	1.00476–1.14253	1.00848	0.95717–1.06254	1.00242	0.97249–1.03327
lag24	1.03225	0.99193–1.07420	1.07185	1.00496–1.14318	1.00926	0.95702–1.06435	1.00296	0.97271–1.03415

Table 3. Air pollution effect on men and women at the Physical Examination Center of the Sichuan Provincial People's Hospital in Chengdu, China (December 2, 2013–December 2, 2016) – cont.

Lag days	Air pollutant							
	PM _{2.5} [$\mu\text{g}/\text{m}^3$]	95% CI	PM ₁₀ [$\mu\text{g}/\text{m}^3$]	95% CI	NO ₂ [$\mu\text{g}/\text{m}^3$]	95% CI	O ₃ [$\mu\text{g}/\text{m}^3$]	95% CI
Men – cont.								
lag25	1.03188	0.99194–1.07344	1.07147	1.00469–1.14269	1.00964	0.95668–1.06553	1.00328	0.97288–1.03463
lag26	1.03118	0.99178–1.07214	1.07024	1.00387–1.14099	1.00957	0.95612–1.06601	1.00336	0.97301–1.03467
lag27	1.03011	0.99142–1.07030	1.06808	1.00241–1.13805	1.00902	0.95526–1.06580	1.00320	0.97307–1.03427
lag28	1.02864	0.99076–1.06797	1.06493	1.00009–1.13396	1.00794	0.95388–1.06506	1.00279	0.97295–1.03354
lag29	1.02676	0.98963–1.06528	1.06072	0.99659–1.12898	1.00628	0.95164–1.06406	1.00211	0.97242–1.03270
lag30	1.02443	0.98777–1.06245	1.05540	0.99141–1.12352	1.00401	0.94802–1.06332	1.00116	0.97110–1.03215
lag31	1.02163	0.98480–1.05985	1.04891	0.98389–1.11823	1.00109	0.94238–1.06345	0.99992	0.96844–1.03242
lag32	1.01835	0.98027–1.05790	1.04121	0.97331–1.11386	0.99747	0.93406–1.06519	0.99839	0.96392–1.03409
lag33	1.01455	0.97376–1.05705	1.03226	0.95900–1.11111	0.99313	0.92251–1.06915	0.99655	0.95712–1.03761
lag34	1.01022	0.96493–1.05764	1.02201	0.94056–1.11050	0.98802	0.90744–1.07576	0.99441	0.94787–1.04323
lag35	1.00534	0.95366–1.05982	1.01044	0.91792–1.11228	0.98212	0.88880–1.08524	0.99194	0.93618–1.05102
Women								
lag0	1.15434	1.06160–1.25519	1.30906	1.12990–1.51664	1.07239	0.91084–1.26260	1.00158	0.89839–1.11663
lag1	1.12269	1.04462–1.20659	1.25259	1.10591–1.41873	1.06666	0.92902–1.22468	1.00624	0.91926–1.10145
lag2	1.09452	1.02719–1.16627	1.20325	1.08084–1.33952	1.06107	0.94269–1.19433	1.00978	0.93623–1.08910
lag3	1.06952	1.00939–1.13324	1.16019	1.05473–1.27620	1.05566	0.95182–1.17082	1.01228	0.94915–1.07960
lag4	1.04741	0.99155–1.10642	1.12271	1.02798–1.22616	1.05042	0.95672–1.15330	1.01381	0.95803–1.07284
lag5	1.02793	0.97418–1.08466	1.09018	1.00142–1.18680	1.04539	0.95808–1.14065	1.01447	0.96322–1.06845
lag6	1.01087	0.95781–1.06687	1.06208	0.97608–1.15565	1.04056	0.95687–1.13158	1.01434	0.96537–1.06579
lag7	0.99601	0.94287–1.05214	1.03794	0.95284–1.13064	1.03597	0.95413–1.12482	1.01351	0.96536–1.06406
lag8	0.98318	0.92964–1.03981	1.01738	0.93228–1.11026	1.03162	0.95074–1.11937	1.01207	0.96398–1.06256
lag9	0.97222	0.91820–1.02941	1.00005	0.91463–1.09344	1.02752	0.94731–1.11452	1.01011	0.96186–1.06079
lag10	0.96297	0.90856–1.02064	0.98563	0.89991–1.07952	1.02370	0.94420–1.10988	1.00773	0.95943–1.05848
lag11	0.95531	0.90065–1.01329	0.97387	0.88798–1.06806	1.02016	0.94159–1.10528	1.00502	0.95691–1.05555
lag12	0.94911	0.89433–1.00723	0.96451	0.87861–1.05881	1.01692	0.93950–1.10071	1.00206	0.95445–1.05205

lag13	0.94425	0.88948-1.00238	0.95735	0.87156-1.05160	1.01399	0.93787-1.09630	0.99896	0.95209-1.04813
lag14	0.94063	0.88595-0.99867	0.95219	0.86654-1.04631	1.01140	0.93656-1.09221	0.99578	0.94982-1.04397
lag15	0.93814	0.88360-0.99605	0.94885	0.86332-1.04286	1.00915	0.93544-1.08866	0.99264	0.94762-1.03979
lag16	0.93670	0.88229-0.99447	0.94717	0.86163-1.04119	1.00725	0.93435-1.08584	0.98960	0.94546-1.03581
lag17	0.93621	0.88190-0.99386	0.94698	0.86129-1.04120	1.00573	0.93319-1.08391	0.98677	0.94330-1.03224
lag18	0.93658	0.88233-0.99416	0.94814	0.86210-1.04278	1.00460	0.93190-1.08296	0.98421	0.94115-1.02924
lag19	0.93773	0.88349-0.99529	0.95052	0.86393-1.04578	1.00387	0.93050-1.08303	0.98203	0.93907-1.02694
lag20	0.93957	0.88532-0.99715	0.95396	0.86668-1.05003	1.00356	0.92906-1.08403	0.98029	0.93715-1.02541
lag21	0.94202	0.88775-0.99961	0.95833	0.87027-1.05531	1.00369	0.92774-1.08585	0.97909	0.93555-1.02466
lag22	0.94500	0.89076-1.00255	0.96351	0.87467-1.06137	1.00427	0.92671-1.08832	0.97851	0.93443-1.02467
lag23	0.94843	0.89430-1.00583	0.96934	0.87982-1.06796	1.00532	0.92616-1.09125	0.97863	0.93399-1.02540
lag24	0.95221	0.89834-1.00932	0.97568	0.88569-1.07483	1.00687	0.92626-1.09449	0.97955	0.93444-1.02682
lag25	0.95628	0.90284-1.01288	0.98241	0.89220-1.08173	1.00892	0.92713-1.09793	0.98134	0.93596-1.02892
lag26	0.96054	0.90772-1.01644	0.98936	0.89924-1.08851	1.01150	0.92883-1.10154	0.98410	0.93867-1.03174
lag27	0.96491	0.91285-1.01994	0.99640	0.90660-1.09508	1.01464	0.93131-1.10542	0.98794	0.94265-1.03539
lag28	0.96929	0.91804-1.02341	1.00335	0.91394-1.10152	1.01835	0.93437-1.10987	0.99293	0.94787-1.04014
lag29	0.97361	0.92298-1.02701	1.01008	0.92073-1.10811	1.02265	0.93758-1.11544	0.99920	0.95412-1.04641
lag30	0.97776	0.92723-1.03104	1.01641	0.92618-1.11542	1.02758	0.94027-1.12299	1.00685	0.96098-1.05491
lag31	0.98166	0.93020-1.03596	1.02217	0.92929-1.12433	1.03315	0.94150-1.13371	1.01601	0.96782-1.06661
lag32	0.98520	0.93119-1.04235	1.02720	0.92889-1.13592	1.03940	0.94025-1.14899	1.02680	0.97386-1.08263
lag33	0.98831	0.92954-1.05079	1.03133	0.92391-1.15124	1.04635	0.93560-1.17021	1.03937	0.97851-1.10403
lag34	0.99087	0.92477-1.06169	1.03438	0.91360-1.17113	1.05404	0.92697-1.19853	1.05388	0.98145-1.13165
lag35	0.99280	0.91667-1.07526	1.03619	0.89767-1.19609	1.06250	0.91416-1.23492	1.07048	0.98265-1.16616

Bolded are significantly effective lag days.

18–65, whose immune functions differed from those of children.

The results presented in this study demonstrate that the effects of air pollution on eosinophils differed between men and women. Many studies have reported that the innate immunity of women, both humoral and cellular, is stronger than that of men [28–30]. Estrogen promotes the development of the T helper 2 (Th2) response and stimulates the secretion of proinflammatory cytokines (IL-4, IL-5 and IL-10) [29]. Of note is the fact that IL-5 can help recruit eosinophils from the blood to inflammatory foci [29,31]. In this study, the eosinophil responses to $PM_{2.5}$ and PM_{10} were different in men and women, as shown in the Figures 1a and 1b, and Figures 1e and 1f. The maximal effect was on the first day of exposure in women, then it gradually decreased within 2–3 weeks after exposure. On the other hand, men were less affected by $PM_{2.5}$ and PM_{10} during the first few days of exposure, but then showed increased effects as the number of lag days increased, reaching maximum values in 3–4 weeks. This study confirms a gender-dependent difference in immune function, with eosinophils in women being more involved than in men.

The authors acknowledge some limitations of their study. First, they did not include the subjects' histories of medication, allergy, pregnancy, menopause, or smoking, all of which may influence eosinophil counts. In addition, occupational and lifestyle factors may lead to some difference in effective exposure to ambient air pollution between men and women. Fortunately, the large cohort of physical examination records, which was used in this study, would reduce the deviation of various factors, enabling the authors to clarify the associations between air pollution, eosinophils, and gender.

CONCLUSIONS

The air pollutants of PM_{10} have a significant effect on human eosinophils for both women and men, but with different temporal patterns, with women showing a lag

of 0–5 days and men showing a lag of 20–28 days. Notably, $PM_{2.5}$ was significant for women with a lag of 0–3 days while it was not significant for men.

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