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We would like to submit the paper “The impact of ambient particulate matter on hospital outpatient visits for respiratory and circulatory system disease in an urban Chinese population” by Ce Wang*, Lan Feng and Kai Chen for publication in Science of the Total Environment.

Significance and Novelty:

There are limited evidence on the association between short-term exposure to ambient particulate matter (PM) and overall hospital outpatient visits for respiratory system disease (RESA) and cardio-cerebrovascular system disease (CCD) in high-polluted countries like China. It is unclear whether this linear exposure-response relationship hold in high pollution area. A time-series study during 2013-2016 was conducted to investigate 245,442 and 430,486 hospital visits for RESA and CCD respectively from Nanjing city, China. The results showed that PM_{2.5} and PM₁₀ were associated with health outcome, and also implied that environmental policies should focus on the multi-pollutants joint prevention and control other than those related to PM only.

We considered this paper based on a huge Chinese dataset, with focusing on a very relevant environmental topic as fitting exactly in your journal's scope. All authors approved the paper, which has not been published previously nor is being considered by another peer-reviewed journal. This manuscript includes 6000 words, five figures and two tables, and supplementary material contains five tables and two figures.

We hope you consider our paper worthwhile of peer review and publication, and are looking forward to your response.

Sincerely,
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- PM was associated with overall outpatient visits of cardiopulmonary health.
- The health impact of air pollution in Nanjing may be due to mixed-pollution.
- The relationship curve presented non-linear across the full range of exposures.
- Estimated risks in warm season were higher than those in cold season.

1 **The impact of ambient particulate matter on hospital**
2 **outpatient visits for respiratory and circulatory system**
3 **disease in an urban Chinese population**

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24 **Abstract**

25 There are limited evidence on the association between short-term exposure to
26 ambient particulate matter (PM) and overall hospital outpatient visits for respiratory
27 system disease (RES) and cardio-cerebrovascular system disease (CCD) in high-
28 polluted countries like China. Though previous epidemiological studies of RES and
29 CCD generally applied a linear relationship of the acute PM effects, it is unclear
30 whether this linear exposure-response relationship hold in high pollution area. In this
31 study, a time-series study during 2013 through 2016 was conducted to investigate
32 245,442 and 430,486 hospital visits for RES and CCD respectively from Nanjing city,
33 China. A combination of logistic generalized additive model (GAM) and distributed
34 lag nonlinear models (DLNM) was used to evaluate the exposure-response associations.
35 The results disclosed that a 10 $\mu\text{g}/\text{m}^3$ increase in $\text{PM}_{2.5}$ and PM_{10} concentration on the
36 current day of exposure (lag 0) was associated with 0.36% (95% CI: -0.02%-0.73%)
37 and 0.33% (0.07%-0.60%) increase in RES; and 0.42% (0.00%-0.85%) and 0.37%
38 (0.08%-0.67%) increase in CCD. The exposure-response association was
39 approximately linear within 0-150 $\mu\text{g}/\text{m}^3$ of PM concentration and non-linear across the
40 full range of exposures. There were no any obvious threshold concentration below
41 which air pollutant had no effect on RES and CCD. The effects of PM on RES and
42 CCD were sensitive to additional adjustment for co-pollutants, indicating the health
43 effects of air pollution mixture in Nanjing city. Though not statistically significant, the
44 estimated risks in warm season were higher than those in cold season, suggesting
45 potential synergistic effects of ambient PM pollution and temperature on triggering

46 RESD and CCD.

47

48 **Keywords:** respiratory; cardiovascular; outpatient visits; air pollution; PM_{2.5}; PM₁₀

49 **1. Introduction**

50 It has been shown that short-term and long-term exposure to ambient particulate
51 matter (PM) are associated with adverse health outcomes for respiratory (e.g., asthma,
52 chronic obstructive pulmonary disease) and cardio-cerebrovascular disease (e.g.,
53 coronary heart disease, stroke), and even mortality (Apte et al., 2015; Burnett et al.,
54 2014; Chen et al., 2017a; Miller et al., 2007; Wang et al., 2018a; Zheng et al., 2018b).
55 Previous epidemiological studies have estimated the relative risk (RR) and excess risk
56 (ER) of short-term PM exposure on respiratory system disease (hereafter referred to
57 RESD) and cardio-cerebrovascular system disease (hereafter referred to CCD) to
58 evaluate population health (Chan et al., 2006; Kim et al., 2012; Tao et al., 2014;
59 Tramuto et al., 2011; Vahedian et al., 2017). These results in general presented
60 “heterogeneity” characteristic, probably due to different species compositions, air
61 pollution sources and population characteristics, etc. For establishing effective control
62 policy, it is necessary to evaluate the local impact of ambient PM pollution on
63 population health, particularly in heavy polluted regions.

64 In general, China was subjected to air pollution since 1950s, and received
65 sufficient attention during two decades (Fang et al., 2016; Liu et al., 2017a). Besides
66 healthy years of life lost, it also caused a large amount of economic costs, e.g.,

67 approximately 25.2 billion USD in health expenditure caused by PM_{2.5} pollution in
68 2030 (Cohen et al., 2017; Xie et al., 2016; Zhang et al., 2008). More regular observation
69 of air pollutants (PM_{2.5}, PM₁₀, SO₂, NO₂, CO, O₃) was implemented since 2013 based
70 on a wide national monitoring network (Liang et al., 2017; Wang et al., 2018c). During
71 2013, China experienced serious air pollution episode and PM was considered as the
72 primary air pollutants. Then, the government issued an *Action Plan on Prevention and*
73 *Control of Air Pollution* which was intended to intensify the comprehensive treatment
74 to reduce the multi-pollutant emissions (State Council, 2013; Chen et al., 2018a).
75 However, till now, very few studies in China had assessed the association between PM
76 concentrations and outpatient visits of overall cardiopulmonary morbidity (Cai et al.,
77 2015; Yang et al., 2016; Zhang et al., 2018; Zhao et al., 2017). Furthermore, before
78 2013, precise estimated impacts of PM pollution exposure were insufficient due to
79 limited availability of air quality observations in China. Different from western
80 countries, regular outpatient visit (first-come first-served) in China greatly
81 outnumbered hospital admission which meant patient's condition was serious, and
82 would broadly reflected the acute health effect associated with air pollution (Guo et al.,
83 2018; Tian et al., 2018b). As a result, it can be considered as a more comprehensive
84 "indicator" to characterize the exposure-response relationship. Hence, it is essential to
85 implement an overall epidemiologic evidence of adverse health effects caused by
86 ambient PM pollution based on a refined air pollution monitoring and health outcome
87 records since 2013.

88 Nanjing city, which is the capital of Jiangsu Province, China, has an area of

89 approximately 6587 km² and the population has increased from 6.43 million in 2013 to
90 6.63 million in 2016 (NMBS, 2014-2017). According to daily average concentration of
91 each pollutant in 2013, day counts for exceeding the Ambient Air Quality Standards
92 (AAQS) Grade II were 310 days for particulate matter with an aerodynamic diameter
93 of 2.5µm or less (PM_{2.5}) (AAQS: 35 µg/m³), 314 days for particulate matter with an
94 aerodynamic diameter between 2.5µm and 10µm (PM₁₀) (AAQS: 70 µg/m³),
95 respectively (MEE, 2016). The overall pollution level was abated in the following years.
96 However, compared with western countries, the city was still subjected to high level of
97 air pollution. In this study, we examined the exposure-response relationship between
98 PM (PM_{2.5} and PM₁₀) and outpatient visit for RESD and CCD across all ages, as well
99 as combined effects of other air pollutants, during 2013 through 2016 in Nanjing City.

100 **2. Methods**

101 **2.1 Data collection**

102 Daily hospital outpatient visit for RESD (all diseases related to respiratory system)
103 and CCD (all diseases related to circulatory system) from 2013 through 2016 were
104 collected from Nanjing Drum Tower Hospital. For some missing historical records on
105 Chinese holidays, (e.g., Spring Festival, Mid-Autumn Festival, National Day), which
106 accounted for 2.3% of the total records, piecewise cubic Hermite interpolation was
107 applied to transform the original records to equidistant data at daily intervals. During
108 weekends, the population could only get access to regular hospital visit till noon,
109 resulting in much lower records on weekends than those on weekdays (see

110 Supplemental Fig.S1 and Fig.S2). Thus, we excluded the visits on weekends in this
111 analysis. The daily concentrations of air pollutants at nine air quality monitoring
112 stations across the city (see Fig.1), i.e., 24-hour average concentration of PM_{2.5}, PM₁₀,
113 SO₂, NO₂, and CO, and the maximum daily 8-hour moving average concentration of
114 O₃, from 2013 to 2016 were obtained from Qingyue Open Environmental Data Center
115 (<https://data.epmap.org>). There are no missing data for daily air pollution
116 concentrations. The matched meteorological observations, i.e., air temperature
117 (ATEMP) and relative humidity (RHUM), were derived from China Meteorological
118 Data Service Center (CMDC) (<http://data.cma.cn>).

119 **2.2 Statistical analysis**

120 Generalized additive model (GAM) (see Equ. (1)) was established to estimate the
121 RR of PM pollution on RESD and CCD with a quasi-Poisson regression, controlling
122 for measured confounders (e.g., RHUM) and unmeasured confounders (e.g., time
123 trends) (Dominici et al., 2002; Peng and Dominici, 2008). Distributed Lag Non-Linear
124 Model (DLNM) was used to evaluate the cumulative effects of air temperature
125 (ATEMP) which considered the non-linear and delayed effect of ATEMP on health
126 (Gasparrini, 2011). The single lag model was specified with daily PM_{2.5} and PM₁₀
127 concentration in various lag days (lag 0-lag 7), and the ER was presented as percent
128 increase with 95% confidence interval (CI) in hospital visit per 10 µg/m³ increase in air
129 pollutant concentration (see Equ. (2)). Then, moving average lag model (lag 01-lag 07)
130 was used to evaluate the cumulative effect of PM pollution within eight days (Arbex et

131 al., 2009; Liu et al., 2013). For evaluating the stability of pollutant effect, the multi-
 132 pollutant model was applied to estimate confounding effect of multiple air pollutants
 133 (Duan et al., 2015; Mostofsky et al., 2012; Zhang et al., 2017). The lag which yielded
 134 the largest effect for the air pollutant of interest in single lag model was applied in the
 135 multi-pollutant model when adjusted for co-pollutant, i.e., SO₂, NO₂, CO and O₃ (Chen
 136 et al., 2016b; Zhu et al., 2018). PM_{2.5} and PM₁₀ were separately input the multi-pollutant
 137 model as they showed highly collinearity (Table S1). We also explored potential effect
 138 of RESD and CCD by cold (October to March) and warm season (April to September),
 139 and tested for important differences between two seasons (Zeka et al., 2006). The
 140 exposure-response curves were illustrated to show the shape of relationship between
 141 PM (PM_{2.5} and PM₁₀) concentrations against hospital visits (RESD and CCD) using
 142 univariate penalized cubic regression spline smooths (degree of freedom, $df = 3$).

$$143 \quad \text{Log} [E(Y_t)] = \alpha + \beta x_{t-l} + \eta \text{Cb.temp}_L + \delta \text{DOW} + s(\text{TIME}, df) + s(\text{RHUM}, df) \quad (1)$$

$$144 \quad \text{ER} = 100 \times \left[e^{IQR \times (\beta_{t-l} \pm 1.96 SE)} - 1 \right] \quad (2)$$

145 $E(Y_t)$ represented the expected count of hospital visits for RESD and CCD at day
 146 t . The pollutant x_{t-l} (e.g., PM_{2.5}) was included in the model at a lag l that might range
 147 from 0 to 7 days lag in single lag model. α was the interception. β represented the
 148 log-relative risk of RESD and CCD associated with a unit increase of air pollutant
 149 concentration. SE showed the standard error of β . Cb.temp_L was cross-basis
 150 function representing an exposure-lag-response bi-dimensional function for PM_{2.5} and
 151 PM₁₀, respectively, and L referred to the maximum lag day. DOW was the dummy
 152 variable for day of week (Monday to Friday) and adjusted as categorical variables.

153 *TIME* was numeric value of 1-1461 (a total of 4 years). Penalized cubic regression
154 splines (*S*) were used to control calendar time (*TIME*) and relative humidity (*RHUM*).
155 η and δ were the coefficients for *Cb.temp_L* and *DOW* respectively. We used 3
156 *df* for *RHUM*, and the cross-basis matrix was generated using a natural cubic spline
157 with 4 *df* for ATEMP and 4 *df* for lag days (Chen et al., 2016a; Liu et al., 2017b;
158 Tian et al., 2017). The *df* for long-term time trend (*df*=6 per year) was selected
159 because of data reduction on hospital visits during weekends (Guo et al., 2018; Guo et
160 al., 2010), and the maximum lag of ATEMP (max *lag* = 14) were chosen in main
161 analysis. We also assessed the robustness of the results in terms of the *df* values for
162 time trend (5 and 7 per year) and maximum lag (7 and 21) (SM Table S2-S3). All
163 analysis of GAM and DLNM were implemented using MGCV package in R-language
164 software version 3.4.4 (R Core Team, 2016). The statistical tests were two-sided and *p*-
165 value < 0.05 was considered as statistically significant. In order to facilitate comparison,
166 the results were presented as the percent change in daily hospital visit per 10 $\mu\text{g}/\text{m}^3$
167 increase of air pollutant concentration.

168 **3. Results**

169 **3.1 Hospital visits and environmental observations**

170 During 2013-2016, the total hospital visits for RESD and CCD were 279,416 and
171 479,397, respectively, and daily records ranged from 35-462 and 31-624 person-time,
172 respectively. The annual averages of PM_{2.5}, PM₁₀ and NO₂ concentrations were still in
173 violation of corresponding AAQS, and the maximum concentration of PM_{2.5}, PM₁₀,

174 SO₂, NO₂, CO and O₃ were 327, 446, 139, 142, 4752 and 280 µg/m³, respectively (see
175 Table 1). Interquartile range (IQR) of PM_{2.5} and PM₁₀ concentration were 47µg/m³ and
176 76.5 µg/m³ during study period. PM_{2.5} concentrations were highly positively associated
177 with PM₁₀ ($r = 0.926, p < 0.05$), and both of them showed moderately positive correlation
178 with SO₂, NO₂ and CO concentrations ($r = 0.634-0.728, p < 0.05$), however, they slightly
179 negatively associated with O₃. In general, air pollutants were negatively correlated with
180 ATEMP and RHUM, except for O₃ concentrations which were directly proportional to
181 ATEMP (see SM Table S1). In general, both hospital visits and environmental variables
182 displayed periodic variations (see Supplemental Fig.S2).

183 **3.2 Exposure-response relationships**

184 Fig.2 illustrated the percent changes for RESD and CCD associated with a 10
185 µg/m³ increase in each air pollutant concentration at different lag structures. In general,
186 after adjustment for calendar time, day of the week and weather conditions, the
187 increments of hospital visits were highest on current day (lag 0) for PM_{2.5} and PM₁₀
188 concentration, i.e., a 10 µg/m³ increase in PM_{2.5} and PM₁₀ was associated with a 0.36%
189 (95% CI: -0.02%, 0.73%) ($p = 0.065$) and 0.33% (0.07%, 0.60%) increase in daily
190 hospital visits on RESD, respectively; 0.42% (0.00%, 0.85%) and 0.37% (0.08%,
191 0.67%) increase in daily hospital visits on CCD, respectively. PM₁₀ also exerted a high
192 effect for RESD (0.33% increase; 95% CI: -0.15%-0.82%) and CCD (0.37%; -0.17%-
193 0.92%) when 7 days moving average exposures (lag 07), however, no significance
194 relationship could be found. The more detailed results were listed in Table S4.

195 We also examined the robustness of exposure-responses relationship with
196 inclusion of potential confounding factors, i.e., SO₂, NO₂, CO and O₃, based on multi-
197 pollutant model (see Table 2). In general, the significant positive associations were still
198 maintained between cardiopulmonary health and PM in multi-pollutant models after
199 adjusted for O₃, and the associations became non-significant after adjusted for SO₂,
200 NO₂ and CO (except for the effect of PM₁₀ on RESD). Specifically, ER of RESD and
201 CCD caused by PM_{2.5} was slightly increased by 0.08% and 0.09% after inclusion of O₃;
202 For PM₁₀, ER of RESD was slightly increased by 0.05% and 0.04% after inclusion of
203 CO and O₃, respectively, and ER on CCD was slightly increased by 0.04% after
204 inclusion of O₃. For the other cases, the estimated effects of PM_{2.5} and PM₁₀ were
205 alleviated and became insignificant with inclusion of corresponding co-pollutants in
206 multi-pollutant models.

207 Fig.3 illustrated the exposure-response relationships associated with each air
208 pollutant exposure (0-99th percentile of concentration ranges of PM_{2.5}: 0-220 µg/m³;
209 PM₁₀: 0-320 µg/m³). The curves shared the similar tendency for RESD and CCD when
210 exposed to the same air pollutant. Approximately, they presented essentially linear
211 relationships within 0-150 µg/m³ of PM concentrations. The curves for PM_{2.5} and PM₁₀
212 tended to become nonlinear at the higher concentration probably due to the data scarcity
213 at this range. The uncertainty of relative risk presented increase at higher concentration
214 of PM (>150 µg/m³). There were no any obvious threshold concentration below which
215 air pollutant had no effect on RESD and CCD.

216 3.3 Effect modification by season

217 It was clearly shown that the association between air pollutants (lag 0 for PM_{2.5}
218 and PM₁₀) and hospital visits presented similarity in cold and warm season (Fig.4).
219 Mostly, the significant positive associations were found, only except for the impact of
220 PM_{2.5} on RESD in cold season. In general, PM had slightly greater impact on RESD
221 and CCD in warm season than in cold season. Particularly, ER of RESD and CCD was
222 increased by 0.18% (0.05%, 0.31%) in cold season and 0.35% (0.20%, 0.50%) in warm
223 season associated with a 10 µg/m³ increase in PM₁₀, respectively. There was no
224 evidence of effect modification by season because of the marginally difference of log-
225 relative risk (see Table S5).

226 4. Discussion

227 This was the first study in China to apply outpatient visits of overall respiratory
228 and circulatory system diseases. From a representative hospital in Nanjing city, short-
229 term exposure to PM was significantly associated with cardiopulmonary morbidity
230 which covered 279,416 and 479,397 outpatient visits for RESD and CCD, respectively.
231 The exposure-response association was approximately linear within 0-150 µg/m³ of PM
232 concentration and non-linear across the full range of exposures. This tendency was
233 similar with the health effect of long-term exposure to air pollution (Burnett et al., 2018).
234 Our findings also demonstrated mixed-pollution in Nanjing city other than PM only. It
235 implied the control of other gaseous pollutants should not be ignored for
236 programmatically protect human health. It should also be noted that apparent health

237 effects of PM were still observed even the concentrations were below the current AAQS
238 of China ($PM_{2.5}$: $35 \mu\text{g}/\text{m}^3$; PM_{10} : $70 \mu\text{g}/\text{m}^3$).

239 Hospital admission was frequently used as morbidity outcome in developed
240 counties. In USA and Europe, several investigations showed the impacts of PM on
241 hospital admission of RESD and CCD from residential population across all ages
242 (Capraz et al., 2017; Granados-Canal et al., 2005; Host et al., 2008; Talbott et al., 2014;
243 Tomaskova et al., 2016; Wordley et al., 1997), e.g., a $10 \mu\text{g}/\text{m}^3$ increase in $PM_{2.5}$
244 concentration was associated with an increase of 1.50%-2.50% and 0.50%-1.20% in
245 RESD and CCD admissions, respectively; while a $10 \mu\text{g}/\text{m}^3$ increase in PM_{10}
246 concentration resulted in an increase of 0.61%-2.40% and 1.24%-2.10% in RESD and
247 CCD admissions, respectively. In general, the estimated epidemiological evidences
248 were relatively higher in western countries compared with results in this study, possibly
249 due to: Firstly, it might involve the different compositions of air pollutants with
250 complex chemical components, which potentially presented spatial heterogeneity (Mo
251 et al., 2018; Qiao et al., 2014); Secondly, the saturation effect and harvest effect might
252 be present when PM pollution maintained a high level (Chen et al., 2017b). In fact, the
253 exposure-response relationships (see Fig.3) displayed non-linear across the full range
254 of exposures, particularly in high polluted region (Pope, 2015); Thirdly, outpatient
255 visits were applied rather than hospital admissions in main analysis. In addition, the
256 inconsistency might also be attributed to the influence of population structure and
257 susceptibility, e.g., heterogeneous in socioeconomic status, educational levels, age
258 distribution (Tony Cox, 2013). We obtained a similar ERs compared with limited

259 researches in China. A case-crossover study across 26 largest cities showed that RESD
260 and CCD admissions with ER of 0.26% (0.22%-0.31%) and 0.23% (0.20%-0.26%)
261 respectively were significantly associated with 10 $\mu\text{g}/\text{m}^3$ increase in $\text{PM}_{2.5}$
262 concentration, and for PM_{10} , the results presented 0.21% (0.17%-0.24%) and 0.15%
263 (0.13%-0.17%), respectively (Liu et al., 2018). ER of outpatient visits for respiratory
264 diseases increased by 0.37% (0.26%-0.48%) with a 10 $\mu\text{g}/\text{m}^3$ increase in $\text{PM}_{2.5}$
265 concentration in Shanghai (patients aged ≥ 15 years) (Wang et al., 2018d). A case in
266 Ningbo observed the adverse effects of $\text{PM}_{2.5}$ on CCD hospital visits, i.e., 10 $\mu\text{g}/\text{m}^3$
267 increase in $\text{PM}_{2.5}$ concentration was significantly associated with an ER of 0.60%
268 (0.00%, 1.10%) (patients aged ≥ 18 years) (Zheng et al., 2018a). For each 10 $\mu\text{g}/\text{m}^3$
269 increase in the $\text{PM}_{2.5}$ concentration, outpatient visits for RESD were increased by 0.53%
270 (0.22%-0.84%) in Lanzhou (Chai et al., 2019).

271 In this study, the single-pollutant models (see Fig.2 and Table S4) disclosed the
272 maximum effect of air pollutants at current-day (lag 0) exposures to $\text{PM}_{2.5}$ and PM_{10} . It
273 indicated that the population in Nanjing city were susceptible to exposures to these
274 pollutants and suffered acute response during very short period. It was consistent with
275 previous studies (Phung et al., 2016; Tian et al., 2018a). E.g., a case study from London
276 demonstrated that the highest association with total cardiovascular diseases was
277 observed with PM_{10} at current day (Atkinson et al., 1999). Some specific-city studies
278 also obtained lagged response to PM but with different maximum lag effects (Luo et
279 al., 2018; Qiu et al., 2018). In general, we found that the risk estimates for PM lost
280 statistical significant after adjusted for co-pollutants in multi-pollutant models except

281 for O₃ (see Table 2), potentially due to its weak correlation with other air pollutants (r
282 = -0.206 ~ -0.053) which resulted in the reduction in the possibility of confounding.
283 Meanwhile, the confidence intervals of RRs were widened, potentially because
284 collinearity among air pollutants and the loss of precision from additional covariates.
285 Particularly, for both RESD and CCD, the adverse effect of PM was greatly influenced
286 after adjusted for SO₂ and NO₂, which was similar with previous study in Wuhan city
287 (Wang et al., 2018b). It was difficult to exactly evaluate the independent effect of PM
288 due to a strong correlation between PM and SO₂, NO₂, CO. These findings implied that
289 air pollution in Nanjing city could be considered as the consequence of “mixed-
290 pollution”. We could infer that multiple air pollutants including other gaseous
291 pollutants appeared most responsible for increased risk on cardiopulmonary health of
292 population in Nanjing city. As a result, the corresponding health effect are indispensable
293 in the future works. Relatively, PM had more serious impact on cardiopulmonary health
294 of population in Nanjing city during warm season than cold season (see Fig.4). Nanjing
295 city is known as one of “four ovens” cities in China, with generally hot summers. It is
296 thus necessary to explore the synergistic effects of air pollution and temperature on
297 triggering RESD and CCD, particular on high temperature days (Chen et al., 2018b;
298 Lee et al., 2018; Sun et al., 2019).

299 The study had several limitations. Firstly, as in most time-series studies, the
300 average daily concentrations of air pollutant across monitoring stations were used for
301 population exposure level. This might result in measurement error because individual
302 exposure depended on many cases, e.g., outdoor activities, location of dwelling.

303 However, this error tends to be non-differential and might result in an underestimation
304 of the PM effects (Chen et al., 2017b). Secondly, the historical records of hospital visits
305 were collected from only one hospital in the city, therefore, and it might affect the
306 generalizability of the epidemiological results. Thirdly, this study only covered four
307 years because the regular monitoring of air pollutants was implemented since 2013.
308 Fourthly, the records on weekends which only covered “half a day” resulted in data
309 discontinuity. Therefore, the main analysis did not include weekends. Finally, more
310 studies on the potential non-linear associations between high-level air pollution and
311 respiratory and circulatory system diseases are needed.

312 **5. Conclusions**

313 This time-series study of cause-specific hospital visits provided an opportunity to
314 determine associations of PM exposures with overall respiratory and circulatory system
315 diseases in Nanjing, China. This is one of the few studies on short-term effects of
316 ambient air pollutants on hospital outpatient visits based on a largescale electronic
317 registry database in China. Multiple air pollutants could be responsible for the adverse
318 health outcomes in Nanjing city. In the future, more comprehensive exposure data is
319 needed to explore non-linear exposure-response associations and synergistic effects.
320 Overall, our findings also implied that, to deal with air pollution mixture, environmental
321 policies should focus on the multi-pollutants joint prevention and control other than
322 those related to PM only.

323 **Conflict of interest**

324 All authors declare that they have no actual or potential competing financial
325 interests.

326

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525 **Figure caption**

526

527 Fig.1 The location of Nanjing City, air quality monitoring stations and Nanjing Drum Tower
528 Hospital (9 stations located in downtown area: S1 - Mai gao qiao, S2 - Cao chang men, S3 - Shan
529 xi lu, S4 - Zhong hua men, S5 - Rui jin lu, S6 - Xuan wu hu, S7 - Pu kou, S8 - Ao ti zhong xin, S9
530 - Xian lin da xue cheng; symbol of red-cross represents the hospital)

531

532 Fig.2 Percent change (95% CI) of hospital admissions on RESD and CCD associated with each IQR
533 increase in air pollutant concentrations with different lag days (red triangle represented single lag
534 effect and square represented moving average lag effect)

535

536 Fig.3 The exposure-response curve of pollutants concentrations and hospital admissions on RESD
537 and CCD ($PM_{2.5}$ (Lag 0), PM_{10} (Lag 0))

538

539 Fig.4 Percent change (95% CI) of RESD and CCD associated with each IQR increase in $PM_{2.5}$ (Lag
540 0) and PM_{10} (Lag 0) concentration by season (cool and warm)

541

542

543 **Table 1**

544 **Table 1 Statistics of hospital visits on RESD, CCD, air pollutants and meteorological observations during 2013 through 2016 in Nanjing City**

Variable	RESD	CCD	PM _{2.5}	PM ₁₀	SO ₂	NO ₂	CO	O ₃	ATEMP	RHUM
Units	persons	persons	µg/m ³	µg/m ³	µg/m ³	µg/m ³	mg/m ³	µg/m ³	°C	%
Min result	35	31	4	8	4	14	0.327	5	-6.7	26
Max result	462	624	327	446	139	142	4.752	280	34.6	97
Mean	235.10	412.34	64.02	110.66	24.27	50.39	1.01	96.25	16.66	72.09
Median	236	425	53	98	20	45	0.916	89	17.8	73
Std dev	62.92	99.70	42.09	63.10	16.04	20.85	0.41	49.30	9.03	13.89
Pct5	130	215.6	17	34	8	24	0.53	28	2.2	48
Pct25	196	367	35	65	14	35	0.741	57	8.7	63
Pct50	236	425	53	98	20	45	0.916	89	17.8	73
Pct75	277.75	480	82	141.5	30	62	1.181	129	23.9	83
Pct95	342	539	145	232	53	91	1.859	190.9	30.89	93
IQR	81.75	113	47	76.5	16	27	0.44	72	15.2	20

545 Note: Std dev represents standard deviation; Pct represents percentile; RESD represents overall respiratory system disease; CCD represents overall cardio-
 546 cerebrovascular system disease; ATEMP represents air temperature; RHUM represents relative humidity.

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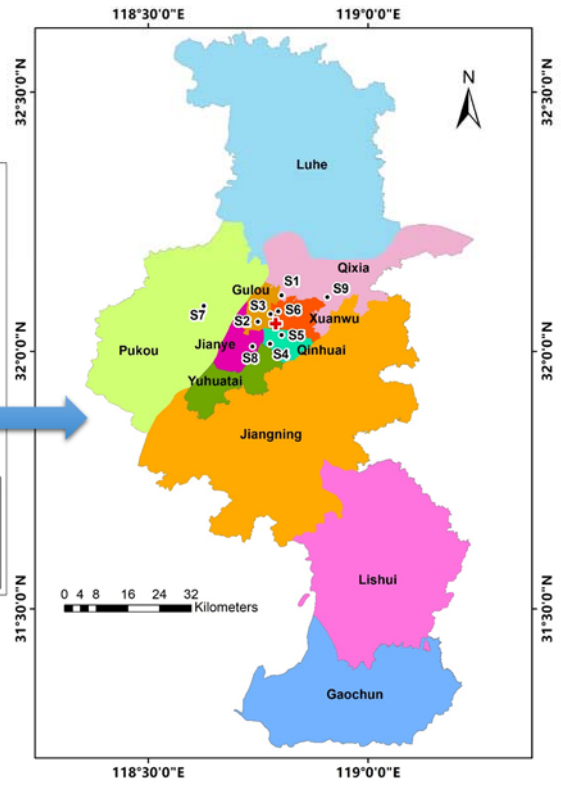
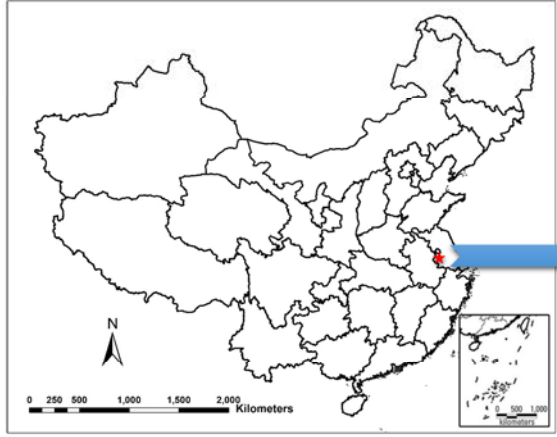
551 **Table 2**

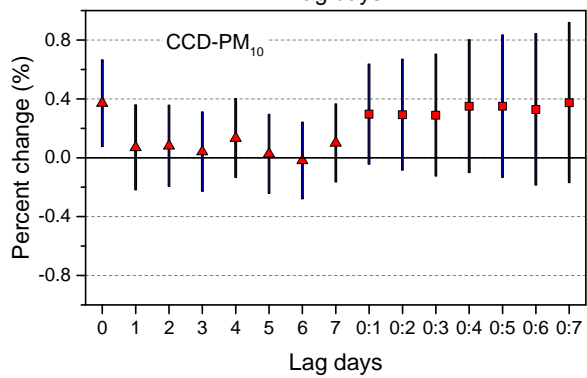
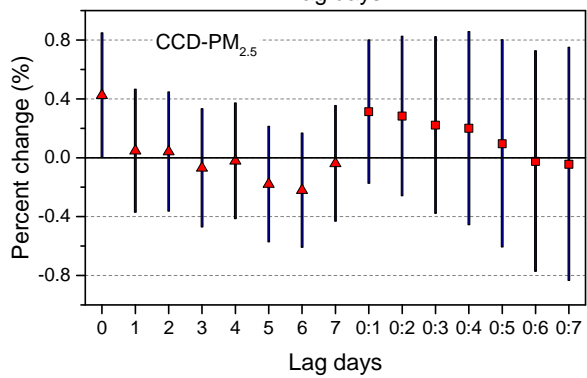
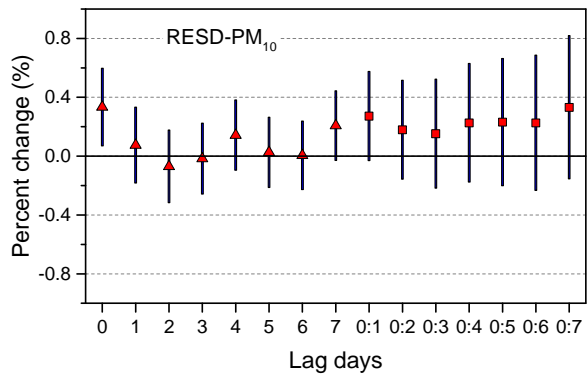
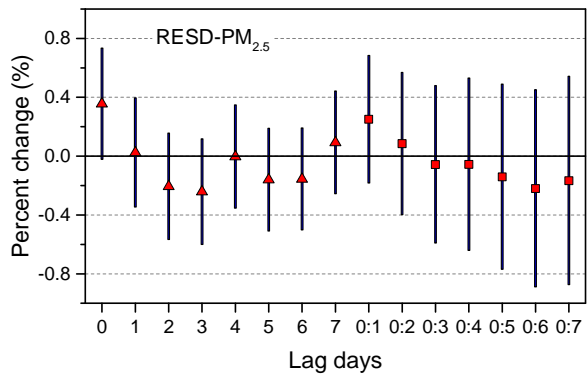
552 **Table 2 ER of RESD and CCD associated with each IQR increase in air pollutant**
 553 **concentrations based on two-pollutant models**

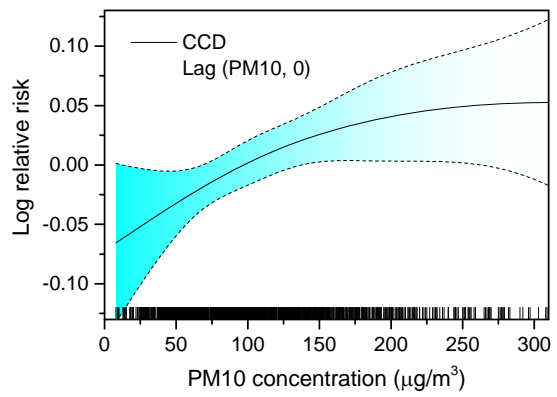
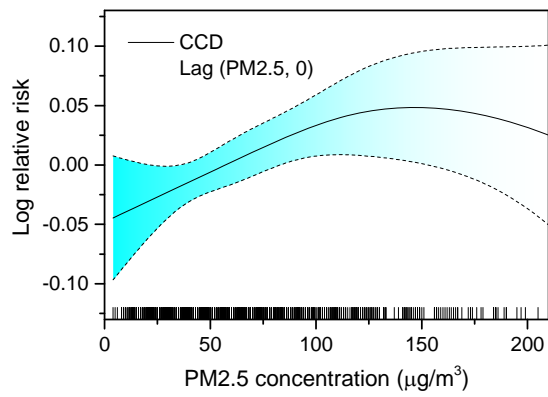
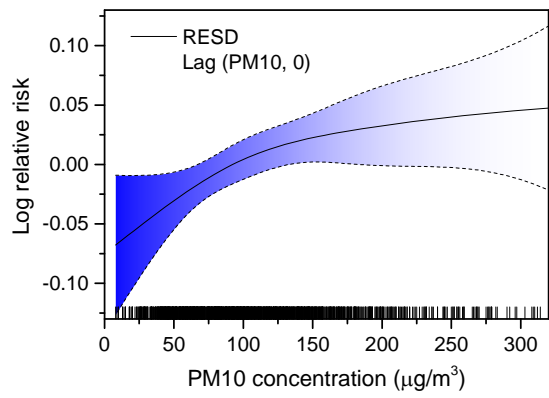
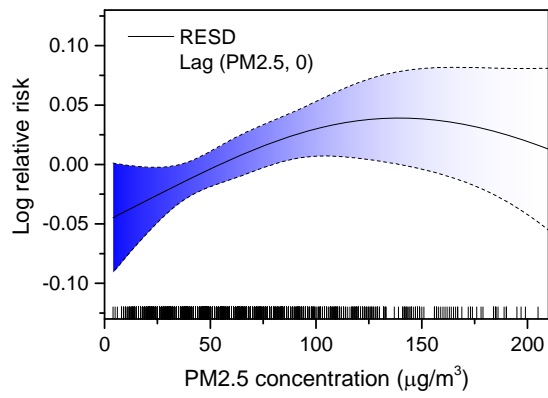
Air pollutants	ER in RESD (%)	95% CI (%)	ER in CCD (%)	95% CI (%)
PM _{2.5}	0.35	-0.02, 0.73	0.42	0.00, 0.85
+SO ₂	-0.03	-0.47, 0.41	-0.16	-0.65, 0.33
+NO ₂	-0.20	-0.64, 0.24	-0.15	-0.64, 0.35
+CO	0.37	-0.15, 0.84	0.23	-0.32, 0.77
+O ₃	0.43	0.05, 0.81	0.51	0.08, 0.94
PM ₁₀	0.33	0.07, 0.60	0.37	0.08, 0.67
+SO ₂	0.08	-0.23, 0.39	-0.03	-0.38, 0.32
+NO ₂	-0.08	-0.40, 0.25	-0.05	-0.41, 0.31
+CO	0.38	0.04, 0.72	0.28	-0.09, 0.65
+O ₃	0.37	0.10, 0.63	0.41	0.11, 0.70

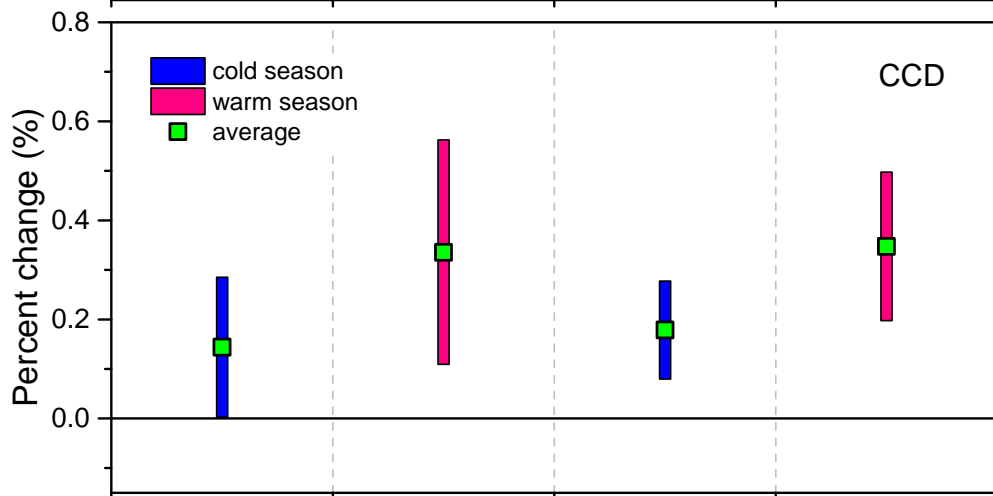
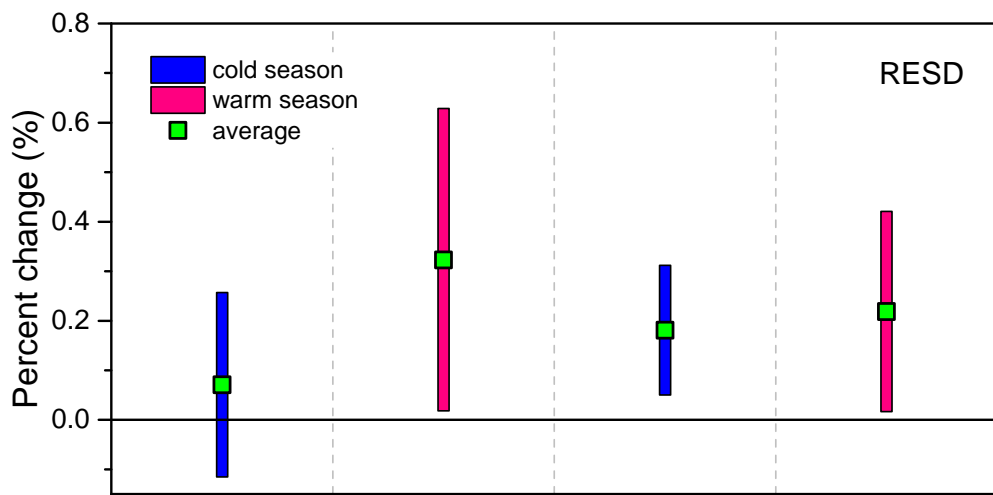
554 Note: in two-pollutant model, PM_{2.5} (Lag 0), PM₁₀ (Lag 0), SO₂ (Lag 0), NO₂ (Lag 0), CO (Lag 0)
 555 and O₃ (Lag 0) concentration was used respectively.

556









PM_{2.5} PM₁₀

Air pollutants

1 **Supplementary Material**

2 **The impact of ambient particulate matter on hospital**
3 **outpatient visits for respiratory and circulatory system disease**
4 **in an urban Chinese population**

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25 **Table S1**

26 **Table S1 Pearson correlations between concentrations of air pollutants and meteorological**
 27 **observations**

Variables	SO ₂	NO ₂	CO	O ₃	PM _{2.5}	PM ₁₀	ATEMP	RHUM
SO ₂	1							
NO ₂	0.651 ^a	1						
CO	0.563 ^a	0.594 ^a	1					
O ₃	-0.100 ^a	-0.166 ^a	-0.206 ^a	1				
PM _{2.5}	0.634 ^a	0.652 ^a	0.709 ^a	-0.116 ^a	1			
PM ₁₀	0.713 ^a	0.728 ^a	0.662 ^a	-0.053 ^a	0.926 ^a	1		
ATEMP	-0.346 ^a	-0.364 ^a	-0.353 ^a	0.494 ^a	-0.338 ^a	-0.320 ^a	1	
RHUM	-0.469 ^a	-0.302 ^a	-0.037	-0.297 ^a	-0.102 ^a	-0.327 ^a	0.134 ^a	1

28 a: two-tailed test of significance is used ($p < 0.05$)

29

30 **Table S2**

31 **Table S2 ER of RESD and CCD caused by PM_{2.5} and PM₁₀ with a maximum lag of 7, 14 and**
 32 **21 days (6 df per year of calendar time was used in DLNM)**

df = 6, per year of calendar time	RESD		CCD	
Maximum lag = 7	ER (%)	95% CI (%)	ER (%)	95% CI (%)
Lag (PM _{2.5} , 0)	0.34	-0.04, 0.72	0.42	-0.00, 0.85
Lag (PM ₁₀ , 0)	0.33	0.07, 0.59	0.39	0.09, 0.68
Maximum lag = 14	ER (%)	95% CI (%)	ER (%)	95% CI (%)
Lag (PM _{2.5} , 0)	0.36	-0.02, 0.73	0.42	0.00, 0.85
Lag (PM ₁₀ , 0)	0.33	0.07, 0.60	0.37	0.08, 0.67
Maximum lag = 21	ER (%)	95% CI (%)	ER (%)	95% CI (%)
Lag (PM _{2.5} , 0)	0.30	-0.08, 0.68	0.38	-0.04, 0.81
Lag (PM ₁₀ , 0)	0.27	0.00, 0.53	0.31	0.01, 0.60

33

34

35

36 **Table S3**

37 **Table S3 ER of RESD and CCD caused by PM_{2.5} by different df per year of calendar time**
 38 **(maximum lag = 14 was used in DLNM)**

df, per year for calendar time	RESD		CCD	
	ER (%)	95% CI (%)	ER (%)	95% CI (%)
df = 5	0.39	0.02, 0.77	0.47	0.05, 0.89
df = 6	0.36	-0.02, 0.73	0.42	0.00, 0.85
df = 7	0.35	-0.04, 0.73	0.43	0.00, 0.86

39 Note: Lag (PM₂₅, 0) was used to display percent increase of hospital visits on RESD and CCD associated
 40 with an IQR increase of PM_{2.5} concentration by different degrees of freedom per year.
 41

42 **Table S4**

43 **Table S4 ER (%) and 95% CI of RESD and CCD for an IQR increase in pollutants concentrations**
 44 **with different lag days in single-pollutant models**

Lag Days	RESD		CCD	
	PM _{2.5}	PM ₁₀	PM _{2.5}	PM ₁₀
Lag 0	0.36 (-0.02, 0.73)	0.33 (0.07, 0.60)	0.42 (0.00, 0.85)	0.37 (0.08, 0.67)
Lag 1	0.02 (-0.35, 0.40)	0.07 (-0.18, 0.33)	0.05 (-0.37, 0.47)	0.07 (-0.22, 0.36)
Lag 2	-0.21 (-0.57, 0.16)	-0.07 (-0.32, 0.18)	0.04 (-0.36, 0.45)	0.08 (-0.20, 0.36)
Lag 3	-0.24 (-0.60, 0.12)	-0.02 (-0.26, 0.22)	-0.07 (-0.47, 0.33)	0.04 (-0.23, 0.31)
Lag 4	0.00 (-0.35, 0.35)	0.14 (-0.10, 0.38)	-0.02 (-0.41, 0.37)	0.13 (-0.13, 0.40)
Lag 5	-0.16 (-0.51, 0.19)	0.03 (-0.21, 0.27)	-0.18 (-0.57, 0.21)	0.03 (-0.24, 0.30)
Lag 6	-0.16 (-0.50, 0.19)	0.01 (-0.23, 0.24)	-0.22 (-0.61, 0.17)	-0.02 (-0.28, 0.24)
Lag 7	0.09 (-0.26, 0.44)	0.21 (-0.03, 0.44)	-0.04 (-0.43, 0.35)	0.10 (-0.16, 0.37)
Lag 01	0.25 (-0.18, 0.68)	0.27 (-0.03, 0.58)	0.31 (-0.17, 0.80)	0.30 (-0.04, 0.64)
Lag 02	0.08 (-0.40, 0.57)	0.18 (-0.16, 0.52)	0.28 (-0.26, 0.83)	0.29 (-0.08, 0.67)
Lag 03	-0.06 (-0.59, 0.48)	0.15 (-0.22, 0.52)	0.22 (-0.38, 0.82)	0.29 (-0.12, 0.70)
Lag 04	-0.06 (-0.64, 0.53)	0.23 (-0.18, 0.63)	0.20 (-0.45, 0.86)	0.35 (-0.10, 0.80)
Lag 05	-0.14 (-0.77, 0.49)	0.23 (-0.20, 0.66)	0.10 (-0.61, 0.80)	0.35 (-0.13, 0.84)
Lag 06	-0.22 (-0.89, 0.45)	0.23 (-0.23, 0.69)	-0.03 (-0.77, 0.73)	0.33 (-0.19, 0.84)
Lag 07	-0.17 (-0.87, 0.54)	0.33 (-0.15, 0.82)	-0.04 (-0.83, 0.75)	0.37 (-0.17, 0.92)

47 **Table S5**48 **Table S5 ER (%) and 95% CI of RESD and CCD for an IQR increase in pollutants concentrations with cold and warm season in single-pollutant models**

Pollutant and statistics	RESD		CCD	
	Cold	Warm	Cold	Warm
PM _{2.5}	0.07 (-0.12, 0.26)	0.32 (0.02, 0.63)	0.14 (0.00, 0.29)	0.34 (0.11, 0.56)
Estimate	7.07×10^{-5}	3.22×10^{-4}	1.44×10^{-4}	3.35×10^{-4}
SE	9.49×10^{-5}	1.55×10^{-4}	7.20×10^{-5}	1.15×10^{-4}
Difference of estimates and 95% CI	-2.52×10^{-4}	$(-3.57 \times 10^{-4}, 3.56 \times 10^{-4})$	-1.91×10^{-4}	$(-2.66 \times 10^{-4}, 2.66 \times 10^{-4})$
PM ₁₀	0.18 (0.05, 0.31)	0.22 (0.02, 0.42)	0.18 (0.08, 0.28)	0.35 (0.20, 0.50)
Estimate	1.81×10^{-4}	2.18×10^{-4}	1.78×10^{-4}	3.47×10^{-4}
SE	6.67×10^{-5}	1.03×10^{-4}	5.03×10^{-5}	7.62×10^{-5}
Difference of estimates and 95% CI	-3.74×10^{-5}	$(-2.40 \times 10^{-4}, 2.40 \times 10^{-4})$	-1.69×10^{-4}	$(-1.79 \times 10^{-4}, 1.79 \times 10^{-4})$

49 Note: in single-pollutant model, PM_{2.5} (Lag 0) and PM₁₀ (Lag 0) concentration was used respectively.

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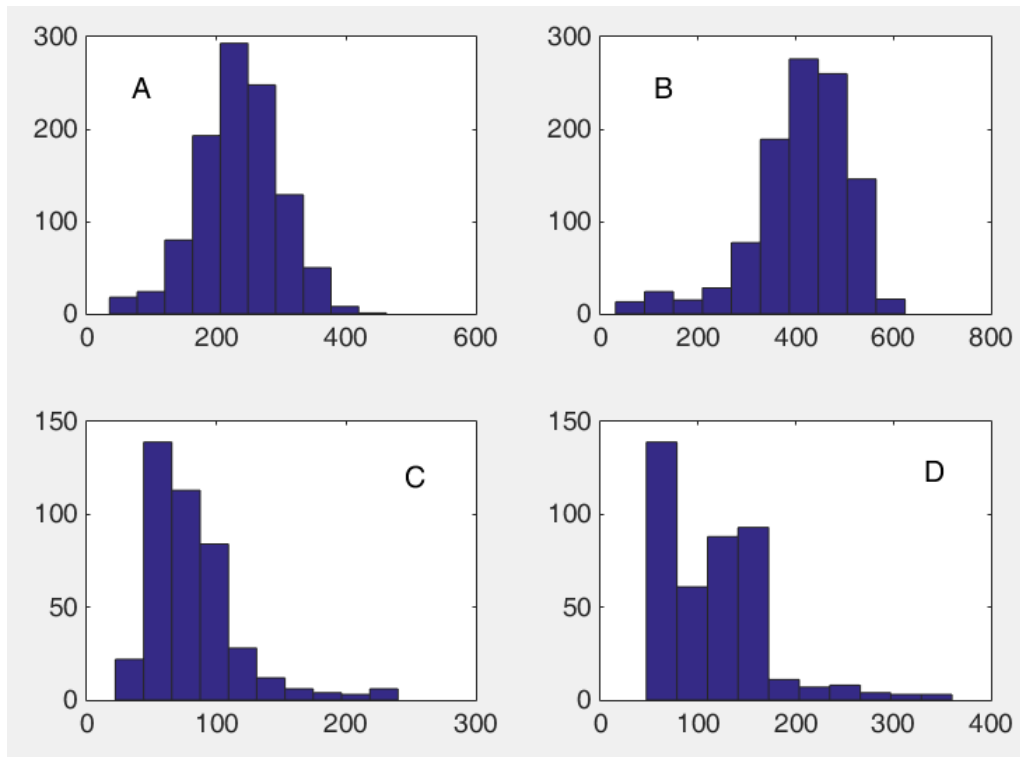
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56 **Figure S1**

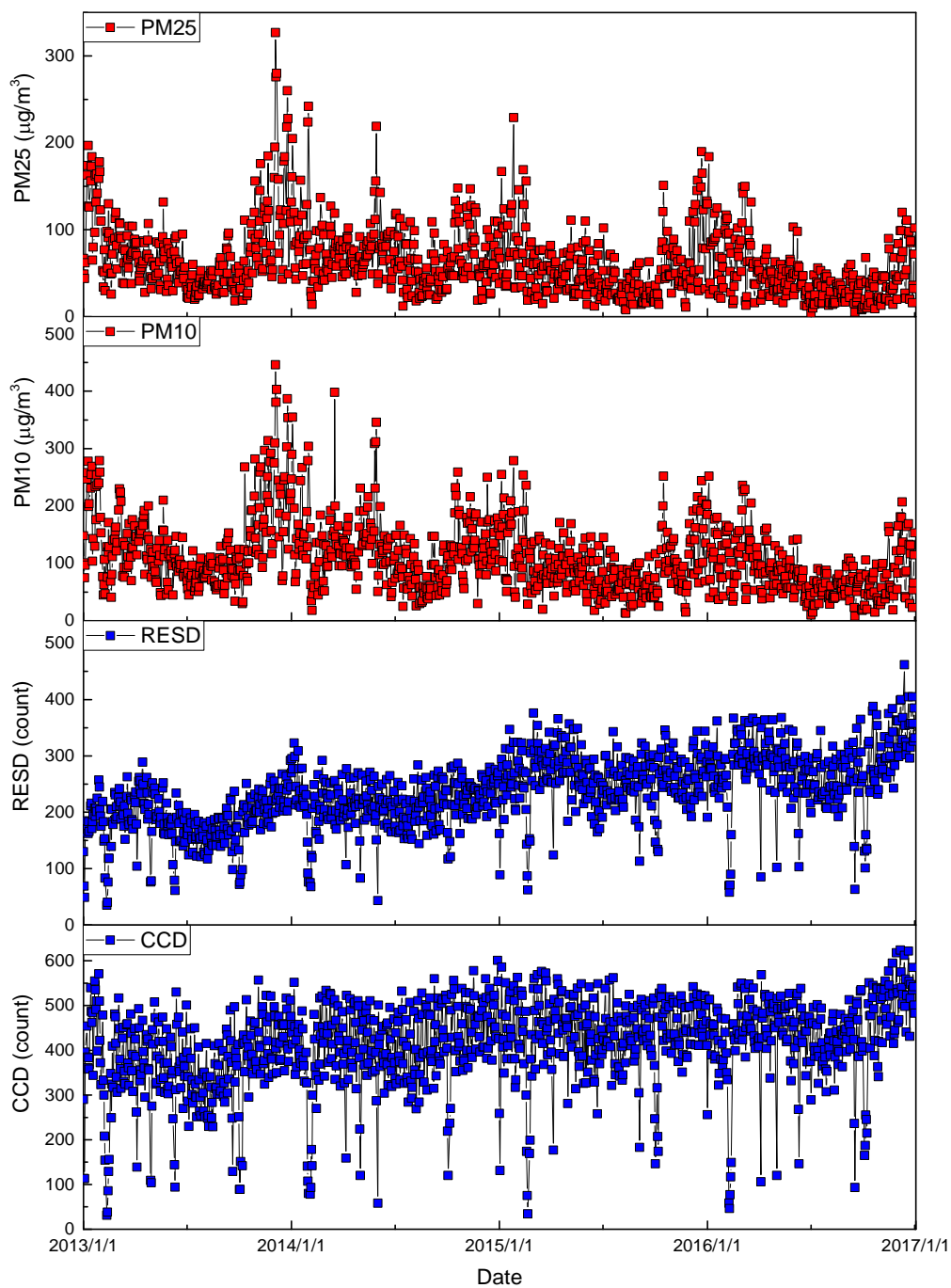


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58 **Fig.S1 Histogram plots for RESD and CCD during 2013 through 2016. (A) RESD on weekdays; (B)**
59 **CCD on weekdays; (C) RESD on weekends; (D) CCD on weekends.**

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61 **Figure S2**



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63 **Fig.S2 Time-series of outpatient visits for RESD and CCD, and concentrations of PM_{2.5} and PM₁₀**
64 **during 2013 through 2016 in Nanjing city**

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