

1 **Exposure to ambient air pollution and blood lipids in adults: the 33 Communities Chinese**  
2 **Health Study**

3 Bo-Yi Yang<sup>a</sup>, Michael S. Bloom<sup>a,b</sup>, Iana Markevych<sup>c</sup>, Zhengmin (Min) Qian<sup>d</sup>, Michael G.  
4 Vaughn<sup>e</sup>, Lenise A. Cummings-Vaughn<sup>f</sup>, Shanshan Li<sup>g</sup>, Gongbo Chen<sup>g</sup>, Gayan Bowatte<sup>h</sup>,  
5 Jennifer L. Perret<sup>i</sup>, Shyamali C. Dharmage<sup>j</sup>, Joachim Heinrich<sup>k</sup>, Steve Hung-Lam Yim<sup>l</sup>, Shao  
6 Lin<sup>b</sup>, Linwei Tian<sup>m</sup>, Mo Yang<sup>a</sup>, Kang-Kang Liu<sup>a</sup>, Xiao-Wen Zeng<sup>a</sup>, Li-Wen Hu<sup>a</sup>,  
7 Yuming Guo<sup>g,\*</sup>, Guang-Hui Dong<sup>a,\*</sup>

8 <sup>a</sup>Guangzhou Key Laboratory of Environmental Pollution and Health Risk Assessment;  
9 Guangdong Provincial Engineering Technology Research Center of Environmental and  
10 Health risk Assessment; Department of Preventive Medicine, School of Public Health, Sun  
11 Yat-sen University, Guangzhou 510080, China.

12 <sup>b</sup>Departments of Environmental Health Sciences and Epidemiology and Biostatistics, University  
13 at Albany, State University of New York, Rensselaer, NY, USA.

14 <sup>c</sup>Institute and Clinic for Occupational, Social and Environmental Medicine, University  
15 Hospital, LMU Munich; Institute of Epidemiology, Helmholtz Zentrum München-German  
16 Research Center for Environmental Health, Neuherberg, Germany.

17 <sup>d</sup>Department of Epidemiology, College for Public Health and Social Justice, Saint Louis  
18 University, Saint Louis 63104, USA.

19 <sup>e</sup>School of Social Work, College for Public Health and Social Justice, Saint Louis University,  
20 Saint Louis 63104, USA.

21 <sup>f</sup>Division of Geriatrics and Nutritional Science, School of Medicine, Washington

22 University-St. Louis, 4921 Parkview Place, St.Louis, MO 63110, USA.

23 <sup>g</sup>Department of Epidemiology and Preventive Medicine, School of Public Health and  
24 Preventive Medicine, Monash University, Melbourne VIC 3004, Australia.

25 <sup>h</sup>Allergy and Lung Health Unit, Centre for Epidemiology and Biostatistics, School of  
26 Population & Global Health, The University of Melbourne, Melbourne, Australia; National  
27 Institute of Fundamental Studies, Kandy, Sri Lanka.

28 <sup>i</sup>Allergy and Lung Health Unit, Centre for Epidemiology and Biostatistics, School of  
29 Population & Global Health, The University of Melbourne, Melbourne, Australia.

30 <sup>j</sup>Allergy and Lung Health Unit, Centre for Epidemiology and Biostatistics, School of  
31 Population & Global Health, The University of Melbourne, Melbourne, Australia; Murdoch  
32 Childrens Research Institute, Melbourne, Australia.

33 <sup>k</sup>Institute and Clinic for Occupational, Social and Environmental Medicine, University  
34 Hospital, LMU Munich; Comprehensive Pneumology Center Munich, German Center for  
35 Lung Research.Ziemssenstrasse 1, 80336 Muenchen, Germany.

36 <sup>l</sup>Department of Geography and Resource Management, The Chinese University of Hong  
37 Kong, Hong Kong, Shatin, N.T., Hong Kong, China; Stanley Ho Big Data Decision  
38 Analytics Research Centre, The Chinese University of Hong Kong, Shatin, N.T.,  
39 Hong Kong, China; Institute of Environment, Energy and Sustainability, The Chinese Uni-  
40 versity of Hong Kong, Shatin, N.T., Hong Kong, China.

41 <sup>m</sup> School of Public Health, Li KaShing Faculty of Medicine, The University of Hong Kong,  
42 Hong Kong Special Administrative Region, China

43 **\* Address correspondence to:**

44 (1) Guang-Hui Dong, MD, PhD, Professor, Guangzhou Key Laboratory of Environmental  
45 Pollution and Health Risk Assessment, Department of Preventive Medicine, School of Public  
46 Health, Sun Yat-sen University, 74 Zhongshan 2<sup>nd</sup> Road, Yuexiu District, Guangzhou 510080,  
47 China. Phone: +862087333409; Fax: +862087330446. E-mail:donggh5@mail.sysu.edu.cn;  
48 donggh512@hotmail.com

49 (2) YumingGuo, MD, PhD, Associate Professor, Department of Epidemiology and Preventive  
50 Medicine, School of Public Health and Preventive Medicine, Monash University, Melbourne,  
51 VIC 3004, Australia. Phone: +61399056100; Fax: +61399056100.E-mail:  
52 yuming.guo@monash.edu

53 **ABSTRACT**

54 **Background:** Little information exists on the lipidemic effects of air pollution, particularly in  
55 developing countries. We aimed to investigate the associations of long-term exposure to  
56 ambient air pollutants with lipid levels and dyslipidemias in China.

57 **Methods:** In 2009, a total of 15,477 participants aged 18-74 years were recruited from the 33  
58 Communities Chinese Health Study conducted in three Northeastern China cities. Total  
59 cholesterol (TC), triglycerides (TG), high-density lipoprotein cholesterol (HDL-C), and  
60 low-density lipoprotein cholesterol (LDL-C) were measured in participants' blood  
61 specimens. Three year (2006-08) average air pollution concentrations were assessed using data  
62 from 33 communities (particles with diameters  $\leq 1.0 \mu\text{m}$  ( $\text{PM}_{10}$ ) and  $\leq 2.5 \mu\text{m}$  ( $\text{PM}_{2.5}$ ) predicted  
63 using a spatial statistical model) or 11 air monitoring stations (particles with diameters  $\leq 10$   
64  $\mu\text{m}$  ( $\text{PM}_{10}$ ), sulfur dioxide ( $\text{SO}_2$ ), nitrogen dioxide ( $\text{NO}_2$ ), and ozone ( $\text{O}_3$ )). Associations were  
65 evaluated by two-level logistic and generalized linear regression models.

66 **Results:** We detected many significant associations between exposure to air pollutants  
67 (especially for  $\text{PM}_{10}$  and  $\text{PM}_{2.5}$ ) and blood lipid levels. Most of the associations suggested  
68 deleterious effects on blood lipid markers (e.g., a  $10 \mu\text{g}/\text{m}^3$  increase in  $\text{PM}_{10}$  was associated  
69 with 1.6% (95% confidence interval (CI): 1.1, 2.0), 2.9% (95% CI: -3.3, 9.3), and 3.2% (95%  
70 CI: 2.6, 3.9) higher levels of TC, TG, and LDL-C, respectively, but 1.4% (95% CI: -1.8, -0.9)  
71 lower HDL-C levels), although beneficial associations were found for  $\text{O}_3$ . In analysis with  
72 dyslipidemias, all the observed associations suggested deleterious lipidemic effects of air  
73 pollutants, and no significant beneficial association was observed for  $\text{O}_3$ . Stratified analyses  
74 showed that the associations were stronger in overweight or obese participants; sex and

75 agemodified the associations, but the pattern of effects was mixed.

76 **Conclusions:** Long-term ambient air pollution was associated with both altered lipid profiles  
77 and dyslipidemias, especially among overweight or obese participants.

78 **Key words:** Particulate matter, Gaseous pollutants, Dyslipidemia, Lipids, Cross-sectional study

79 **Abbreviations**

80 BMI, body mass index; CI, confidence interval; CVD, cardiovascular diseases; HDL-C,  
81 high-density lipoprotein cholesterol; LDL-C, low-density lipoprotein cholesterol; NO<sub>2</sub>,  
82 nitrogen dioxide; OR, odds ratio; O<sub>3</sub>, ozone; PM<sub>1</sub>, particles with diameters ≤ 1.0 μm; PM<sub>2.5</sub>,  
83 particles with diameters ≤ 2.5 μm; PM<sub>10</sub>, particles with diameters ≤ 10 μm; PM<sub>2.5-10</sub>, particles  
84 with diameters ranging from 2.5 to 10 μm; SEPA, the State Environmental Protection  
85 Administration of China; SO<sub>2</sub>, sulfur dioxide; TC, total cholesterol; TG, triglycerides;  
86 33CCHS, the 33 Chinese Community Health Study.

87

88 **1. Introduction**

89 Cardiovascular diseases(CVD)are responsible for approximately 31% of deaths worldwide  
90 (World Health Organization, 2017) and the results of numerous epidemiological studies have  
91 supported a causal relation for long-term air pollution exposure withCVD(Brook et al., 2010;  
92 Bourdrel et al., 2017).Inhaled air pollutants triggerinflammation, oxidative stress, autonomic  
93 imbalance, and epigenetic changes (Brook et al., 2010; Bourdrel et al., 2017).These reactions  
94 have been linked to several CVD risk factors, includingatherosclerosis, hypertension, diabetes  
95 mellitus, and dyslipidemia(Brook et al., 2010;Thiering and Heinrich, 2015; Rajagopalan and  
96 Brook, 2012; Yang et al., 2018). For example, higher levels of particulate matter (PM) have  
97 been linked to increased systemic inflammation (Brook et al. 2010), which can lead to adverse  
98 lipid metabolism and lipid oxidation (Chen et al., 2013). The relationshipsof air pollution with  
99 hypertension and diabetes mellitus have been demonstrated bymany epidemiological and  
100 experimental studies(Brook et al., 2010; Rajagopalan and Brook, 2012; Thiering and Heinrich,  
101 2015), including our own(Dong et al., 2013; Yang et al., 2017; Yang et al., 2018).

102 Dyslipidemia, characterized by hypercholesterolemia, hypertriglyceridemia,  
103 hypoalphalipoproteinemia, and/orhyperbetalipoproteinemia, is the foremost cause of  
104 atherosclerosis(Stensland-Bugge et al., 2000) and is inextricably relatedto the development of  
105 CVD(Expert Panel on Detection, Evaluation, and Treatment of High Blood Cholesterol in  
106 Adults,2001).The global prevalence of dyslipidemias is high and increasing (Cahalin et al.,  
107 2014). For example, a 2013 report from the American Heart Association suggested that the  
108 prevalences of hypercholesterolemia, hypoalphalipoproteinemia, and  
109 hyperbetalipoproteinemia in American adults were 43.4%, 21.8%, and 31.1%, respectively

110 (Go et al., 2013). Several previous epidemiological studies explored relationships between  
111 ambient air pollutant exposure and dyslipidemia and blood lipid levels, yet the results were  
112 inconsistent (Bell et al., 2017; Bind et al., 2016; Cai et al., 2017; Chuang et al., 2011; Jiang et  
113 al., 2016; Poursafa et al., 2014; Shanley et al., 2016; Sørensen et al., 2015; Wallwork et al.,  
114 2017; Yeatts et al., 2007; Yitshak Sade et al., 2016) (see Table S1). Additionally, most studies  
115 investigated effects among specific populations, including asthmatics (Yeatts et al., 2007),  
116 patients with chronic diseases (Yitshak Sade et al., 2016), the elderly (Bind et al., 2016;  
117 Chuang et al., 2011; Sørensen et al., 2015; Wallwork et al., 2017), and adolescents (Poursafa  
118 et al., 2014). However, the lipidemic effects of air pollution exposure were rarely evaluated in  
119 general populations. Moreover, previous studies were mostly conducted in high-income  
120 nations or regions (Bell et al., 2017; Bind et al., 2016; Chuang et al., 2011; Shanley et al.,  
121 2016; Sørensen et al., 2015; Wallwork et al., 2017; Yeatts et al., 2007; Yitshak Sade et al.,  
122 2016). There are few data available to characterize the risks of air pollution exposure on lipid  
123 levels and dyslipidemia in low-income countries.

124 In recent decades, China has experienced a gradual increase in the prevalence of dyslipidemia,  
125 although it remains lower than that in many developed countries (Pan et al., 2016). For instance,  
126 the 2002 China National Nutrition and Health Survey reported that the prevalences of  
127 hypercholesterolemia, hyperbeta lipoproteinemia, hypoalphalipoproteinemia, and  
128 hypertriglyceridemia were 2.9%, 2.5%, 7.4%, and 11.9%, respectively (Zhao, 2008), while the  
129 corresponding 2013-2014 China Chronic Disease and Risk Factor Surveillance  
130 prevalences were 6.9%, 8.1%, 20.4%, and 13.8% (Zhang et al. 2018). Simultaneously, air  
131 pollution has emerged as a severe environmental problem in China (Guan et al., 2016; Rohde



132 and Muller, 2015). Given temporal increases in both ambient air pollution and the prevalence  
133 of dyslipidemia, and the scarcity of data, it is of significant public health importance to  
134 explore the relationship between the two. To begin to address the data gap, this study  
135 examined associations between long-term residential ambient air pollution and blood lipid levels  
136 in a large community-based sample of urban adults participating in the 33 Chinese Community  
137 Health Study (33CCHS).

## 138 **2. Methods**

### 139 **2.1. Study population**

140 The population of the 33CCHS was previously described in detail (Dong et al., 2013; Yang et  
141 al., 2017). Briefly, in 2009, we used a random-number generator coupled to a four-staged,  
142 stratified, cluster sampling strategy to recruit study participants. First, to maximize the  
143 inter-city gradients of air pollutants, we randomly selected three cities - Shenyang, Anshan,  
144 and Jinzhou - from 14 total cities in Liaoning province. There are five districts in Shenyang  
145 city and three each in the cities of Anshan and Jinzhou. Second, we randomly selected three  
146 communities from each of the districts, generating a total of 33 study communities. Each  
147 study community was approximately 0.25-0.64 km<sup>2</sup> in area. Third, we randomly selected  
148 700-1000 households from each study community. Fourth, from each study household, we  
149 randomly selected one adult aged 18 to 74 years for study enrollment. To be included,  
150 individuals had to live at the study address for at least five years, have no severe pre-existing  
151 diseases (e.g., cancers), and not be pregnant. Based on the sampling frame, 28,830 participants  
152 were invited, of whom 24,845 individuals completed the survey, yielding an overall response  
153 rate of 86.2%. A total of 9368 individuals were excluded from the present analysis due to

154 refusal to provide a blood sample, leaving a final sample of 15,477 participants (62.3% of the  
155 33CCHS participants). All participants completed informed consent prior to study enrollment,  
156 and Sun Yat-Sen University's Human Studies Committee reviewed and approved all study  
157 procedures and protocols.

## 158 **2.2. Health outcomes**

159 After an overnight fast, peripheral venous blood samples were collected from study  
160 participants. Total cholesterol (TC), triglycerides (TG), high-density lipoprotein cholesterol  
161 (HDL-C), and low-density lipoprotein cholesterol (LDL-C) levels were determined using a  
162 Hitachi Autoanalyzer (Type 7170A; Hitachi Ltd.; Tokyo, Japan). Hypercholesterolemia was  
163 defined as  $TC \geq 240$  mg/dL; hypertriglyceridemia was defined as  $TG \geq 200$  mg/dL;  
164 hypoalphalipoproteinemia was defined as  $HDL-C \leq 40$  mg/dL; and hyperbetalipoproteinemia  
165 was defined as  $LDL-C \geq 160$  mg/dL (Joint Committee for Developing Chinese Guidelines on  
166 Prevention and Treatment of Dyslipidemia in Adults, 2007).

## 167 **2.3. Air pollution data**

168 A detailed description of the exposure assessment was provided in our previous  
169 publications (Dong et al., 2013; Yang et al., 2017; Chen et al., 2018). Briefly, daily  
170 concentrations of PM with diameters  $\leq 1.0$   $\mu\text{m}$  ( $PM_{1.0}$ ) and  $\leq 2.5$   $\mu\text{m}$  ( $PM_{2.5}$ ) were predicted for  
171 the 33 study communities, at a  $0.1^\circ \times 0.1^\circ$  spatial resolution, using  $PM_{1.0}$  and  $PM_{2.5}$   
172 measurements from air monitoring stations, satellite remote sensing, meteorology, and land  
173 use characteristics. Aerosol optical depth data was combined from two types of Moderate  
174 Resolution Imaging Spectroradiometer algorithms—Dark Target and Deep Blue. A

175 generalized additive model was developed to link ground-monitored PM<sub>1</sub> and PM<sub>2.5</sub> data with  
176 aerosol optical depth data and other spatial and temporal predictors. In each of the study  
177 districts, there was one air monitoring station, which was located within a 1-km distance from  
178 the centroid of the community of each study participant's home address (Fig. S1). We  
179 collected data for PM with diameters  $\leq 10 \mu\text{m}$  (PM<sub>10</sub>), sulfur dioxide (SO<sub>2</sub>), nitrogen dioxide  
180 (NO<sub>2</sub>) and ozone (O<sub>3</sub>) concentrations from 11 air monitoring stations, according  
181 to standardized procedures set by the State Environmental Protection Administration of China  
182 (SEPA) (SEPA, 1992). These air monitoring stations were mandated to be away from main  
183 traffic roads, industry sources, or residential sources of emissions from the combustion of coal,  
184 waste, or oil. Thus, air pollution measurements from these stations were more likely to reflect  
185 mixtures from different sources and background levels in urban areas. Daily average  
186 concentrations of PM<sub>10</sub>, SO<sub>2</sub>, NO<sub>2</sub>, and an eight-hour average of O<sub>3</sub> were calculated using  
187 measurements from days with at least 75.0% of one-hour values available. PM<sub>coarse</sub> (PM<sub>2.5-10</sub>)  
188 was calculated by subtracting PM<sub>2.5</sub> from PM<sub>10</sub>. The three-year (2006-08) average  
189 concentrations of air pollutants were calculated for the 33 communities and 11 districts (or air  
190 monitoring stations), and then assigned to each participant based on proximity of the  
191 community or district to his/her residential address, as surrogates of long-term air pollution  
192 exposure.

#### 193 **2.4. Covariates**

194 All potential covariates were selected *a priori*. An interviewer administered questionnaire was  
195 used to collect information on age (years), sex (male/female), nationality (Han/others),  
196 household annual income ( $\leq 5000$  Yuan/5001-10000 Yuan/10001-30000 Yuan/ $\geq 30000$  Yuan),

197 highest educational attainment (no school/primary school/middle school/junior college or  
198 higher), current smoking (yes/no), alcohol drinking (yes/no), regular exercise (yes/no),  
199 controlled diet with low calories and low fat (yes/no), sugar-sweetened soft drink  
200 consumption ( $\leq 1$  day per week/ 2-4 days per week/  $\geq 5$  days per week), family history of  
201 dyslipidemia, and district (or community). Height (cm) and weight (kg) were measured  
202 without heavy clothes and shoes, and body mass index (BMI,  $\text{kg}/\text{m}^2$ ) was calculated. Data on  
203 temperature, humidity, and wind speed for the 11 districts were obtained from Liaoning  
204 Provincial Meteorological Bureau. Per-capita gross domestic product and population density  
205 in each district were obtained from Shenyang, Jinzhou, and Anshan cities' Statistical  
206 Yearbooks. Season at the time of blood sampling was also included as a potential cofounder.

## 207 **2.5. Statistical analysis**

208 The Shapiro-Wilk and the Bartlett tests were used to examine data normality and  
209 homogeneity, respectively. Differences in basic characteristics between men and women, and  
210 between participants with and without blood sampling, were tested using Student's t-test, or  
211 the Wilcoxon rank sum or chi-square tests. We used the Spearman rank correlation test to  
212 assess the relationship between air pollutants.

213 We applied linear regression models to assess associations between individual air pollutants  
214 (per  $10 \mu\text{g}/\text{m}^3$  increase) and blood lipid levels (TC, TG, HDL-C, and LDL-C), which were  
215 naturally log-transformed to achieve normal distributions. Effect estimates were then  
216 back-transformed from the log scale using  $100 \times [\exp(\beta) - 1]$  and presented as percent  
217 differences with corresponding 95% confidence intervals (CI). We also used two-level binary

218 logistic regression models to investigate associations of ambient air pollutants (per  $10\mu\text{g}/\text{m}^3$   
219 increase) with hypercholesterolemia, hypertriglyceridemia, hypoalphalipoproteinemia and  
220 hyperbetalipoproteinemia, where participants were treated as first-level units and districts or  
221 communities as second-level units. Detailed descriptions of the two-stage binary logistic  
222 regression models were provided in our previous publications (Dong et al., 2013; Yang et al.,  
223 2017) and in the supplemental material (detailed information on two-level binary logistic  
224 regression model). These results are presented as odds ratio (OR) with corresponding 95% CI.  
225 All regression models were adjusted for the variables listed in the Covariates section. District  
226 or community was incorporated as a random effect, and the remaining covariates were  
227 incorporated as fixed effects. Multi-pollutant models could not be applied, as all air pollutants  
228 were correlated moderately to highly (except  $\text{NO}_2$  and  $\text{SO}_2$ ). Therefore, highly correlated  
229 pollutants were regressed against each other and the residuals were then incorporated into the  
230 models for associations between air pollutants and lipid outcomes (Flexeder et al., 2017).

231 We performed subgroup analyses by sex (men, women), age group ( $\geq 50$  years,  $< 50$  years),  
232 and BMI category ( $\geq 25\text{ kg}/\text{m}^2$ ,  $< 25\text{ kg}/\text{m}^2$ ), and across-product term was incorporated into  
233 regression models to evaluate the statistical significance of their interactions. Additionally, we  
234 repeated the regression analyses of air pollution and lipid levels by excluding participants who  
235 took lipid lowering drugs, or those had cardiovascular diseases or diabetes mellitus. Also, we  
236 applied multi-annual average concentrations of air pollutants (i.e., one-year average  
237 (2008), two-year average (2007-08), and three-year average (2006-08)) as long-term exposures,  
238 in order to exclude possible exposure fluctuations over shorter periods. Furthermore, 30-day  
239 average air pollutant concentrations before the blood draw were additionally adjusted to

240 explore the potential impact of short-term air pollution exposure on lipids. Similarly, for  
241 pollutants in which long-term and short-term levels were highly correlated, these were  
242 regressed against each other and the individual residuals were then incorporated into the  
243 regression models.

244 Data analysis was performed using SAS 9.4(SAS Institute, Cary, NC) with a p value less than  
245 0.05 considered as statistically significant for a two-tailed test.

### 246 **3. Results**

#### 247 ***3.1. Descriptive statistics***

248 The mean age of study participants was 45.0 years, and 52.7% were men (Table 1). Most  
249 participants had a middle school or higher education (84.8%). Thirty percent and 24.6% of  
250 them were smokers and drinkers, respectively. Approximately 31.9% reported engaging in  
251 regular exercise and 7.9% had a family history of dyslipidemia. The prevalences of  
252 hypercholesterolemia, hypertriglyceridemia, hypoalphalipoproteinemia, and  
253 hyperbetalipoproteinemia were 11.1%, 22.6%, 18.3%, and 8.6%, respectively. The median  
254 concentrations of TC, TG, HDL-C, and LDL-C were 179.92 mg/dL, 118.58 mg/dL, 50.97  
255 mg/dL, and 98.60 mg/dL, respectively. Men and women differed for all sociodemographic and  
256 lifestyle variables, with the exception of nationality. The distribution of the main  
257 characteristics was similar between the analytical sample and those who were excluded from  
258 this analysis (Table S2)

259 The PM<sub>1</sub>, PM<sub>2.5</sub>, PM<sub>10</sub>, SO<sub>2</sub>, NO<sub>2</sub>, and O<sub>3</sub> concentrations varied greatly across study  
260 districts (or communities) with a median of 62 µg/m<sup>3</sup>, 73 µg/m<sup>3</sup>, 123 µg/m<sup>3</sup>, 48 µg/m<sup>3</sup>, 33 µg/m<sup>3</sup>,

261 and  $50\mu\text{g}/\text{m}^3$ , respectively (Table 2). Except for  $\text{SO}_2$  and  $\text{NO}_2$ , moderate to high  
262 correlations between air pollutants were detected (Spearman correlations ranged from 0.45 to  
263 0.99). In particular,  $\text{NO}_2$  and  $\text{O}_3$  were positively correlated, which might be caused by high air  
264 pollutants levels, especially  $\text{NO}_x$  and volatile organic compounds (Shi et al., 2015; Zong et al.  
265 2017), at the study site.

### 266 ***3.2. Associations between air pollutants and lipid levels***

267 The associations of air pollutants with blood lipid levels are summarized in Tables 3 and S3.  
268 For all participants, higher concentrations of  $\text{PM}_{10}$  and  $\text{PM}_{2.5}$  consistently showed significant  
269 associations with higher levels of TC, TG, and LDL-C, as well as with lower levels of HDL-C.  
270 However, associations of  $\text{PM}_{10}$  with blood lipids were less consistent, and  $\text{PM}_{2.5-10}$  was  
271 associated only with TG levels (Table S3). Higher  $\text{NO}_2$  levels were significantly associated  
272 with higher levels of TC and TG and with lower levels of HDL-C.  $\text{SO}_2$  was positively  
273 associated with TG, but not with the remaining lipid markers. Higher  $\text{O}_3$  concentrations were  
274 significantly associated with higher levels of TG and HDL-C, but with lower levels of TC and  
275 LDL-C. The results remained materially unchanged in sensitivity analyses where 263  
276 individuals taking lipid lowering drugs (Table S4) or 2222 individuals with CVD or diabetes  
277 mellitus (Table S5) were excluded, where multi-annual average concentrations of air pollutants  
278 were used (Table S6), and where the models were additionally adjusted for short-term air  
279 pollutant levels (Table S7). The only exception in the latter case was the statistical  
280 insignificance of the association between  $\text{O}_3$  and HDL-C (Table S7).

281 We detected statistically significant interactions between air pollutant concentrations and sex

282 on blood lipids; however, the pattern of effects was mixed in stratified analyses (Fig. 1; Table  
283 S8). For example, while associations of all six air pollutants with HDL-C were stronger for  
284 men, in the case of TG they were stronger for women. In addition, the positive associations of  
285 LDL-C with PM<sub>1</sub> and PM<sub>2.5</sub> were stronger in women than in men, and were stronger yet  
286 negative for PM<sub>10</sub>, SO<sub>2</sub>, NO<sub>2</sub> and O<sub>3</sub>. In stratified analysis by age, the associations between air  
287 pollutants and lipid levels were similarly complex, although most interaction terms were not  
288 statistically significant (Table S8). In another stratified analysis, BMI significantly modified  
289 the associations of PM<sub>1</sub> and PM<sub>2.5</sub> with HDL-C and LDL-C, with stronger associations among  
290 overweight/obese participants (Table S8).

### 291 *3.3. Associations between air pollutants and dyslipidemias*

292 For all participants, we detected statistically significant associations between: (1) higher PM<sub>1</sub>,  
293 and PM<sub>2.5</sub> concentrations with higher odds for  
294 hypercholesterolemia, hypoalphalipoproteinemia, and hyperbetalipoproteinemia; (2) higher  
295 PM<sub>10</sub>, SO<sub>2</sub>, and O<sub>3</sub> concentrations with higher odds for hypertriglyceridemia; and (3) higher  
296 NO<sub>2</sub> concentrations with higher odds for hypercholesterolemia (Table 4). In stratified analyses  
297 by sex and age, we detected statistically significant interactions of air pollutant concentrations  
298 with sex and age for several dyslipidemia associations, but the pattern was mixed (Table S9). In  
299 stratified analyses by BMI, associations of air pollutants with dyslipidemias (particularly for  
300 hypoalphalipoproteinemia) were consistently greater in participants who were overweight or  
301 obese (Table S9).

## 302 **4. Discussion**



303 **4.1. Key findings**

304 To our knowledge, this is the largest population-based epidemiological study to date, to  
305 explore associations between ambient air pollution and blood lipids in a developing country. We  
306 detected many statistically significant associations between exposure to long-term ambient air  
307 pollutants (particularly PM<sub>1</sub> and PM<sub>2.5</sub>) and blood lipid levels and the prevalence of  
308 dyslipidemias. Most of the associations suggested deleterious effects on blood lipid levels (i.e.,  
309 associated with higher levels of TC, TG, and LDL-C, and lower HDL-C levels), though  
310 beneficial associations were detected for O<sub>3</sub>. However, all observed associations suggested  
311 deleterious effects of air pollutants on dyslipidemias; no beneficial significant association was  
312 observed for O<sub>3</sub> with dyslipidemia. Stratified analyses showed that associations between air  
313 pollutants and lipids were generally stronger among overweight and obese participants; sex  
314 and age also modified associations, but the pattern of effects was complicated.

315 **4.2. Comparison with other studies and interpretations**

316 Several previous studies investigated associations between air pollution exposure and blood  
317 lipid levels or dyslipidemias, but the findings have been inconsistent (Bell et al., 2017; Bind et  
318 al., 2016; Cai et al., 2017; Chuang et al., 2011; Jiang et al., 2016; Poursafa et al., 2014;  
319 Shanley et al., 2016; Sørensen et al., 2015; Wallwork et al., 2017; Yeatts et al., 2007; Yitshak  
320 Sade et al., 2016). Consistent with our current findings, a cross-sectional study from the USA  
321 revealed that an 11.1 µg/m<sup>3</sup> increase in PM<sub>10</sub> was associated with 2.42% higher TG  
322 levels (Shanley et al., 2016). Another cross-sectional study, from Denmark, reported positive  
323 association for PM<sub>2.5</sub> exposure with TC levels. Similarly, a retrospective cohort study of Israeli  
324 adults reported a statistically significant association between higher PM<sub>2.5</sub> concentrations and

325 lower HDL-C levels(Yitshak Sade et al., 2016).Chuang et al.(2011) found thatPM<sub>10</sub> and  
326 NO<sub>2</sub>were significantly associated with higher TC levels in Taiwan, but there were no  
327 associations with TGor HDL-C levels. In addition, a panel study among American adults  
328 showed no significant association between PM<sub>2.5</sub>exposure and TC levels (Yeatts et al., 2007).  
329 The specific reasons for the inconsistent results across studies of ambient air pollution  
330 exposure and blood lipid levels are not clear. They may be related to the differences in  
331 population characteristics (e.g., age, genetic background, lifestyles, and health status) and/or  
332 local or regional differences in the physical and chemical properties of the air pollutants (e.g.,  
333 concentrations, chemical constituents, and sources).

334 The biological mechanisms underlying links between air pollutants and lipid metabolism are  
335 not fully understood. However, several possible biological pathways have been proposed. One  
336 hypothesis is that inhaled air pollution elicits systemic inflammation and oxidativestress  
337 (Lodovici and Bigagli, 2011; Shanley et al., 2016; Thompson et al., 2010), which can induce  
338 adverse lipid metabolism and lipid oxidation (Chen et al., 2013). Air pollutants could also  
339 cause aberrant DNA methylation by decreasing activity of DNA methyltransferases. Several  
340 studies have linked air pollution exposure to abnormalities in global DNA methylation as well  
341 asto methylation at specific genes related to lipid metabolism (Chen et al., 2016; Bind et al.,  
342 2014).The associations between air pollutant concentrations, PM in particular,and blood lipid  
343 levels in our current study are consistent with these hypothesized biologicalmechanisms.

344 In stratified analyses, we found stronger associations between air pollutants and blood lipids  
345 among overweight and obese participants, which are partially consistent with Sørensen et al's

346 work. (2015). Existing evidence shows that both air pollution exposure and  
347 overweight/obesity are associated with higher systemic inflammation (Bastard et al., 2006;  
348 Rajagopalan and Brook, 2012).Overweight and obese participants might therefore be more  
349 vulnerable to adverse health effects from air pollution, which act in part through an  
350 inflammation pathway. We also found that sex and age modified the effects of air pollution on  
351 lipid levels and dyslipidemias, but the pattern was mixed. To thebest of our knowledge, only  
352 two prior studies (Shanley et al., 2016; Sørensen et al., 2015)investigated sex- and  
353 age-specific associations between air pollutants and blood lipids. In line with our linear  
354 regression findings, Shanley et al. (2016) reported that age did not significantly modify the  
355 associations of PM<sub>10</sub> with TC and TG. However, they also observed a stronger association  
356 between PM<sub>10</sub> and TC among women, whereas for PM<sub>10</sub> and TG a stronger association was  
357 observedamong men, which contradicts our results. In another study, Sørensen et al. (2015)  
358 observed that both age and sex did not modify the associations of NO<sub>2</sub> and PM<sub>2.5</sub> with  
359 TC.Collectively, there is limited and inconsistent epidemiological evidenceto characterize the  
360 role of age and sex in modifying air pollution-lipid associations at present, and so further  
361 investigation is merited.

#### 362 ***4.3. Implications for policy makers***

363 Dyslipidemias are well-documented risk factors for CVD (Zhang et al., 2003).Randomized  
364 trials have shown that lipid-lowering treatment could significantly decrease the risk of CVD  
365 (Fulcher et al., 2015). For example, the Asia Pacific Cohort Studies Collaboration reported 35%  
366 and 25% increased risksfor coronary death and incident stroke in Asians, respectively, per  
367 1-mmol/L higher serum TC (Zhang et al. 2003).Yet, a meta-analysis of 22 trials found that a

368 1-mmol/L LDL-C reduction could decrease major CVD events by 21% (Fulcher et al., 2015).  
369 In the current study, we found that a 10- $\mu\text{g}/\text{m}^3$  increase in air pollutants was associated with an  
370 approximately 1-2% increase in blood lipids levels. Although the observed difference was  
371 relatively small and thus of uncertain clinical impact, our findings have certain public health  
372 implications for helping policy makers to develop intervention policies, given the high levels  
373 of air pollution (Guan et al., 2016) and high prevalence of CVD in China (Peters et al. 2017).

#### 374 ***4.4. Strengths and limitations***

375 This study has several strengths. First, our analysis was based on a large sample of  
376 northeastern Chinese with a high response rate, using standardized protocols and instruments,  
377 which ensured sufficient statistical power to detect modest effects, and generalizability of our  
378 results. Second, unlike most previous studies, which focused on specific populations, our  
379 present study provides valuable evidence regarding a general population in a developing  
380 nation. Third, in addition to measuring exposure to the traditional ambient ‘criteria’ air  
381 pollutants (i.e., PM<sub>2.5</sub>, PM<sub>10</sub>, SO<sub>2</sub>, NO<sub>2</sub>, and O<sub>3</sub>), we, for the first time, report effects for PM<sub>1</sub>  
382 on blood lipids. In addition, all air pollutants levels are high in our study settings, thus our  
383 results can provide a valuable reference for other developing countries, such as India. Finally,  
384 a combination of objectively measured lipid levels coupled to a rich set of covariate data  
385 allowed for a comprehensive data analysis, including adjustment for BMI, physical activity,  
386 diet, and meteorological variables to minimize the impact of confounding.

387 Despite the novel nature of our results, several limitations should also be acknowledged. First,  
388 the cross-sectional study design precluded assessment of temporality, and we are thus unable

389 to infer a causal association between air pollution exposure and blood lipid  
390 levels. Second, exposure levels were assigned using data from the nearest air monitor or  
391 community rather than using personal air pollution exposure data, which means that only 33  
392 (for PM<sub>1</sub> and PM<sub>2.5</sub>) and 11 (for PM<sub>10</sub>, NO<sub>2</sub>, SO<sub>2</sub>, and O<sub>3</sub>) unique air pollution values were  
393 available for the 15,477 participants. The values may have misclassified some participants, by  
394 randomly underestimating exposure in some and overestimating exposure in  
395 others. Nevertheless, such exposure misclassification is likely to bias the results towards null  
396 (Hutcheon et al., 2010). This indicates that if we had individual-level data on air pollution  
397 levels, our estimated effects of air pollution on blood lipids would have been stronger than the  
398 current estimates. Furthermore, our exposure assessment did not capture specific emissions  
399 known to adversely affect health, such as traffic-related sources, that are likely to show large  
400 variation across both space and time. Third, participants' baseline characteristics differed  
401 significantly among the study districts and communities (Yang et al., 2017). Although we  
402 collected rich covariate data to adjust for confounding, it is possible that the observed  
403 significant associations were biased by unmeasured confounding factors that differed across  
404 the study districts or communities, including health-care access, available green-space, noise,  
405 and household environments. Unfortunately, these data were not collected by the 33CCHS. In  
406 addition, conditions such as acute infection and inflammation, hyperthyroidism, and nephrotic  
407 syndrome, may affect lipid status (Nigam, 2011). These data were also not available in our  
408 current study, which may have compromised our estimates. Fourth, 62.3% of the 33CCHS  
409 participants with blood specimens were included in the current analysis, and so a selection  
410 bias was not impossible. However, the distribution of baseline characteristics was similar

411 between participants with and without a blood specimen, and so any effect is likely to be  
412 modest. Fifth, we used a questionnaire to collect self-reported information on demographic  
413 and lifestyle characteristics; thus, recall bias and misclassification is possible. Finally,  
414 correlations between air pollutants were generally moderate or high, which limited our ability  
415 to assess the health effects of multiple pollutants simultaneously. However, we performed  
416 regression analyses on air pollutants that highly correlated with each other, and then adjusted  
417 the individual residuals in order to accommodate the co-exposures.

## 418 **5. Conclusions**

419 Our findings suggest that long-term exposure to ambient air pollution is associated with  
420 altered lipid levels and the prevalence of dyslipidemias, especially among overweight and  
421 obese people. However, considering the limitations of our study, future well-designed  
422 longitudinal studies are warranted to more definitively evaluate the effects of ambient air  
423 pollution on lipid metabolism.

## 424 **Declaration of interests**

425 None

## 426 **Acknowledgements**

427 The authors acknowledge the cooperation of participants in this study who have been very  
428 generous with their time and assistance.

## 429 **Funding**

430 The research was funded by the National Natural Science Foundation of China (No.91543208;

431 No.81703179; No.81673128); the National Key Research and Development Program of  
432 China (No.2016YFC0207000); the Fundamental Research Funds for the Central Universities  
433 (No.16ykzd02; No.17ykpy16); theGuangdongProvince Natural Science Foundation  
434 (No.2016A030313342; 2017A050501062); and Science and Technology Program of  
435 Guangzhou (201807010032; 201803010054).YG was supported by the Career Development  
436 Fellowship of AustralianNational Health and Medical Research Council (No.APP1107107).

437 **References**

- 438 Bastard, J.P., Maachi, M., Lagathu, C., Kim, M.J., Carnon, M., Vidal, H., et al., 2006. Recent  
439 advances in the relationship between obesity, inflammation, and insulin resistance. *Eur.*  
440 *Cytokine Netw.* 17, 4-12.
- 441 Bell, G., Mora, S., Greenland, P., Tsai, M., Gill, E., Kaufman, J.D., 2017. Association of air  
442 pollution exposures with high-density lipoprotein cholesterol and particle number: the  
443 Multi-Ethnic Study of Atherosclerosis. *Arterioscler.Thromb.Vasc.Biol.* 37, 976–982.
- 444 Bind, M.A., Lepeule, J., Zanobetti, A., Gasparrini, A., Baccarelli, A., Coull, B.A., et al.,  
445 2014. Air pollution and gene-specific methylation in the normative aging study.  
446 *Epigenetics* 9, 448-458.
- 447 Bind, M.A., Peters, A., Koutrakis, P., Coull, B., Vokonas, P., Schwartz, J., 2016. Quantile  
448 regression analysis of the distributional effects of air pollution on blood pressure, heart  
449 rate variability, blood lipids, and biomarkers of inflammation in elderly American men:  
450 The Normative Aging Study. *Environ.Health Perspect.* 124, 1189-1198.
- 451 Bourdrel, T., Bind, M.A., Bęgot, Y., Morel, O., Argacha, J.F., 2017. Cardiovascular effects of  
452 air pollution. *Arch.Cardiovasc. Dis.* 110, 634-642.
- 453 Brook, R.D., Rajagopalan, S., Pope, C.R., Brook, J.R., Bhatnagar, A., Diez-Roux, A.V., et al.,  
454 2010. Particulate matter air pollution and cardiovascular disease: an update to the  
455 scientific statement from the American Heart Association. *Circulation* 121, 2331-2378.
- 456 Cahalin, L.P., Myers, J., Kaminsky, L., Briggs, P., Forman, D.E., Patel, M.J., et al. 2014.



457 Current trends in reducing cardiovascular risk factors in the United States: focus on  
458 worksite health and wellness. *Prog.Cardiovasc. Dis.* 56, 476-483.

459 Cai, Y., Hansell, A.L., Blangiardo, M., Burton, P.R., BioSHaRE, deHoogh, K., Doiron, D., et  
460 al., 2017. Long-term exposure to road traffic noise, ambient air pollution, and  
461 cardiovascular risk factors in the HUNT and lifelines cohorts. *Eur. Heart J.* 38,  
462 2290-2296.

463 Chen, G., Knibbs, L.D., Zhang, W., Li, S., Cao, W., Guo, J., et al., 2018. Estimating  
464 spatiotemporal distribution of PM<sub>1</sub> concentrations in China with satellite remote sensing,  
465 meteorology, and land use information. *Environ.Pollut.* 233, 1086-1094.

466 Chen, R., Meng, X., Zhang, A., Wang, C., Yang, C., Li, H., et al., 2016. DNA methylation  
467 and its mediation in the effects of fine particulate air pollution on cardiovascular  
468 biomarkers: a randomized crossover trial. *Environ. Int.* 94, 614-619.

469 Chen, T., Jia, G., Wei, Y., Li, J., 2013. Beijing ambient particle exposure accelerates  
470 atherosclerosis in ApoE knockout mice. *Toxicol.Lett.*, 223, 146-153.

471 Chuang, K.J., Yan, Y.H., Chiu, S.Y., Cheng, T.J., 2011. Long-term air pollution exposure and  
472 risk factors for cardiovascular diseases among the elderly in Taiwan. *Occup. Environ.*  
473 *Med.* 68, 64-68.

474 Dong, G.H., Qian, Z.M., Xaverius, P.K., Trevathan, E., Maalouf, S., Parker, J., et al.,  
475 2013. Association between long-term air pollution and increased blood pressure and

476 hypertension in China. *Hypertension* 61, 578-584.

477 Expert Panel on Detection, Evaluation, and Treatment of High Blood Cholesterol in Adults., 2  
478 001. Executive summary of The Third Report of The National Cholesterol Education Pro  
479 gram (NCEP) Expert Panel on Detection, Evaluation, And Treatment of High Blood Cho  
480 lesterol In Adults (Adult Treatment Panel III). *JAMA*. 285, 2486-2497.

481 Flexeder, C., Thiering, E., Koletzko, S., Berderl, D., Lehmann, I., von Berg, A., et al., 2017.  
482 Higher serum 25(OH)D concentrations are associated with improved FEV<sub>1</sub> and FVC in  
483 adolescence. *Eur. Respir. J.* 49, 1601804.

484 Fulcher, J., O'Connell, R., Voysey, M., Emberson, J., Blackwell, L., Mihaylova, B., et al.,  
485 2015. Efficacy and safety of LDL-lowering therapy among men and women:  
486 meta-analysis of individual data from 174,000 participants in 27 randomised trials.  
487 *Lancet* 385, 1397-1405.

488 Go, A.S., Mozaffarian, D., Roger, V.L., Benjamin, E.J., Berry, J.D., Borden, W.B., et al.,  
489 2013. Heart disease and stroke statistics--2013 update: a report from the American  
490 Heart Association. *Circulation* 127, e6-e245.

491 Guan, W.J., Zheng, X.Y., Chung, K.F., Zhong, N.S., 2016. Impact of air pollution on the  
492 burden of chronic respiratory diseases in China: time for urgent action. *Lancet* 388,  
493 1939-1951.

494 Hutcheon, J.A., Chioloro, A., Hanley, J.A., 2010. Random measurement error and regression

495 dilution bias. *BMJ* 340, c2289.

496 Jiang, S., Bo, L., Gong, C., Du, X., Kan, H., Xie, Y., et al., 2016. Traffic-related air pollution is  
497 associated with cardio-metabolic biomarkers in general residents. *Int. Arch. Occup.*  
498 *Environ. Health* 89, 911-921.

499 Joint Committee for Developing Chinese Guidelines on Prevention and Treatment of  
500 Dyslipidemia in Adults., 2007. Chinese guidelines on prevention and treatment of  
501 dyslipidemia in adults. *Chin. J. Epidemiol.* 35, 390-419.

502 Lodovici, M., Bigagli, E., 2011. Oxidative stress and air pollution exposure. *J. Toxicol.* 2011,  
503 487074.

504 Nigam, P.K., 2011. Serum lipid profile: fasting or non-fasting? *Indian J. Clin. Biochem.* 26,  
505 96-97.

506 Pan, L., Yang, Z., Wu, Y., Yin, R.X., Liao, Y., Wang, J., et al., 2016. The prevalence,  
507 awareness, treatment and control of dyslipidemia among adults in China. *Atherosclerosis*  
508 248, 2-9.

509 Peters, S.A., Yang, L., Guo, Y., Chen, Y., Bian, Z., Millwood, I.Y., et al., 2017. Parenthood  
510 and the risk of cardiovascular diseases among 0.5 million men and women: findings  
511 from the China Kadoorie Biobank. *Int. J. Epidemiol.* 46, 180-189.

512 Poursafa, P., Mansourian, M., Motlagh, M.E., Ardalan, G., Kelishadi, R., 2014. Is air quality  
513 index associated with cardiometabolic risk factors in adolescents? The CASPIAN-III

514 Study. Environ. Res. 134, 105-109.

515 Rajagopalan, S., Brook, R.D., 2012. Air pollution and type 2 diabetes: mechanistic insights.  
516 Diabetes 61, 3037-3045.

517 Rohde, R.A., Muller, R.A., 2015. Air pollution in China: mapping of concentrations and  
518 sources. PLoS One 10, e135749.

519 Shanley, R.P., Hayes, R.B., Cromar, K.R., Ito, K., Gordon, T., Ahn, J., 2016. Particulate air  
520 pollution and clinical cardiovascular disease risk factors. Epidemiology 27, 291-298.

521 Shi, J., Deng, H., Bai, Z., Kong, S., Wang, X., Hao, J., et al., 2015. Emission and profile  
522 characteristic of volatile organic compounds emitted from coke production, iron smelt,  
523 heating station and power plant in Liaoning Province, China. Sci. Total Environ.  
524 515-516, 101-108.

525 Sørensen, M., Hjortebjerg, D., Eriksen, K.T., Ketzel, M., Tjønneland, A., Overvad, K., et al.,  
526 2015. Exposure to long-term air pollution and road traffic noise in relation to cholesterol:  
527 A cross-sectional study. Environ. Int. 85, 238-243.

528 State Environmental Protection Administration of China (SEPA), 1992.  
529 Standardized environmental monitoring and analysis methods (Beijing, China).

530 Stensland-Bugge, E., Børnaa, K.H., Joakimsen, O., Njøstad, I., 2000. Sex differences in the  
531 relationship of risk factors to subclinical carotid atherosclerosis measured 15 years later:  
532 the Tromsø Study. Stroke 31, 574-581.

533 Thiering, E., Heinrich, J., 2015. Epidemiology of air pollution and diabetes. *Trends*  
534 *Endocrinol. Metab.* 26, 384-394.

535 Thompson, A.M., Zanobetti, A., Silverman, F., Schwartz, J., Coull, B., Urch, B., et al.,  
536 2010. Baseline repeated measures from controlled human exposure studies: associations  
537 between ambient air pollution exposure and the systemic inflammatory biomarkers IL-6  
538 and fibrinogen. *Environ. Health Perspect.* 118, 120-124.

539 Wallwork, R.S., Colicino, E., Zhong, J., Kloog, I., Coull, B.A., Vokonas, P., et al.,  
540 2017. Ambient fine particulate matter, outdoor temperature, and risk of metabolic  
541 syndrome. *Am. J. Epidemiol.* 185, 30-39.

542 World Health Organization. Cardiovascular diseases (CVDs) Fact Sheet. WHO. Available at:  
543 <http://www.who.int/mediacentre/factsheets/fs317/en/>. Updated May 2017. (accessed Jan  
544 27, 2018).

545 Yang, B.Y., Qian, Z., Howard, S.W., Vaughn, M.G., Fan, S.J., Liu, K.K., et al., 2018. Global  
546 association between air pollution and blood pressure: a systematic review and  
547 meta-analysis. *Environ. Pollut.* 235, 576-588.

548 Yang, B.Y., Qian, Z.M., Vaughn, M.G., Nelson, E.J., Dharmage, S.C., Heinrich, J., et al.,  
549 2017. Is prehypertension more strongly associated with long-term ambient air pollution  
550 exposure than hypertension? Findings from the 33 Communities Chinese Health Study.  
551 *Environ. Pollut.* 229, 696-704.

552 Yeatts, K., Svendsen, E., Creason, J., Alexis, N., Herbst, M., Scott, J., et al., 2007. Coarse  
553 particulate matter (PM<sub>2.5-10</sub>) affects heart rate variability, blood lipids, and circulating  
554 eosinophils in adults with asthma. *Environ. Health Perspect.* 115, 709-714.

555 Yitshak Sade, M., Kloog, I., Liberty, I.F., Schwartz, J., Novack, V., 2016. The association  
556 between air pollution exposure and glucose and lipids levels. *J. Clin. Endocrinol. Metab.*  
557 101, 2460-2467.

558 Zhang, M., Deng, Q., Wang, L., Huang, Z., Zhou, M., Li, Y., et al., 2018. Prevalence of  
559 dyslipidemia and achievement of low-density lipoprotein cholesterol targets in Chinese  
560 adults: a nationally representative survey of 163,641 adults. *Int. J. Cardiol.* 260, 196-203.

561 Zhang, X., Patel, A., Horibe, H., Wu, Z., Barzi, F., Rodgers, A., et al. 2003. Cholesterol,  
562 coronary heart disease, and stroke in the Asia Pacific region. *Int. J. Epidemiol.* 32,  
563 563-572.

564 Zhao, W.H., 2008. The 2002 China National Nutrition and Health Survey, 7, People's  
565 Medical Publishing House, Lipid. Beijing, China (in Chinese).

566 Zong, Z., Wang, X., Tian, C., Chen, Y., Fang, Y., Zhang, F., et al., 2017. First assessment of  
567 NO<sub>x</sub> sources at a regional background site in north China using isotopic analysis linked  
568 with modeling. *Environ. Sci. Technol.* 51, 5923-5931.

569 **Figure legend**

570 **Fig. 1.** Associations between air pollutants and blood lipid levels (A: total cholesterol, TC; B:  
571 triglycerides, TG; C: high-density lipoprotein cholesterol, HDL-C; D: low-density lipoprotein  
572 cholesterol, LDL-C) by sex. Stars represent statistically significant interactions.

573 **Table 1** Study population characteristics (n = 15,477).

Characteristics	Value (mean $\pm$ SD, n(%), or median (Q1, Q3))		
	Total (n=15,477)	Men (n=8156)	Women (n=7321)
Age (years) <sup>a</sup>	44.97 $\pm$ 13.45	44.44 $\pm$ 14.20	45.56 $\pm$ 12.55
Ethnicity			
Han	14,554 (94.0%)	7670 (94.0%)	6884 (94.0%)
Other	923 (6.0%)	486 (6.0%)	437 (6.0%)
Education <sup>a</sup>			
Junior college or higher	3579 (23.1%)	2250 (27.6%)	1329 (18.2%)
Middle school	9554 (61.7%)	5008 (61.4%)	4546 (62.1%)
Primary school	1863 (12.0%)	782 (9.6%)	1081 (14.8%)
No school	481 (3.1%)	116 (1.4%)	365 (5.0%)
Annual family income <sup>a</sup>			
$\leq$ 5000 Yuan	1167 (7.5%)	618 (7.6%)	549 (7.5%)
5001-10,000 Yuan	1977 (12.8%)	846 (10.4%)	1131 (15.5%)
10,001-30,000 Yuan	7869 (50.8%)	4198 (51.5%)	3671 (50.1%)
$\geq$ 30,000 Yuan	4464 (28.8%)	2494 (30.6%)	1970 (26.9%)
Tobacco smoking status <sup>a</sup>			
Nonsmoker	10,837 (70.0%)	4004 (49.1%)	6833 (93.3%)
Smoker	4640 (30.0%)	4152 (50.9%)	488 (6.7%)
Alcohol drinking status <sup>a</sup>			
Nondrinker	11,668 (75.4%)	4562 (55.9%)	7106 (97.1%)
Drinker	3809 (24.6%)	3594 (44.1%)	215 (2.9%)
Regular exercise <sup>a</sup>			
Yes	4932 (31.9%)	2724 (33.4%)	2208 (30.2%)
No	10,545 (68.1%)	5432 (66.6%)	5113 (69.8%)
Control diet with low calorie and fat <sup>a</sup>			
Yes	3861 (24.9%)	1828 (22.4%)	2033 (27.8%)
No	11,616 (75.1%)	6328 (77.6%)	5288 (72.2%)
Sugar-sweetened soft drink intake <sup>a</sup>			
$\leq$ 1 day per week	13,621 (88.0%)	6996 (85.8%)	6625 (90.5%)
2-4 days per week	1286 (8.3%)	818 (10.0%)	468 (6.4%)
$\geq$ 5 days per week	570 (3.7%)	342 (4.2%)	228 (3.1%)
BMI <sup>a</sup>			
$\geq$ 25 kg/m <sup>2</sup>	6271 (40.5%)	3660 (44.9%)	2611 (35.7%)
$<$ 25 kg/m <sup>2</sup>	9206 (59.5%)	4496 (55.1%)	4710 (64.3%)
Family history of dyslipidemia <sup>a</sup>			
Yes	1228 (7.9%)	492 (6.0%)	736 (10.1%)
No	14,249 (92.1%)	7664 (94.0%)	6585 (89.9%)
Blood lipids& dyslipidemias			
TC (mg/dL)	179.92 (155.98, 205.41)	180.70 (157.53, 204.83)	179.15 (154.44, 206.18)
TG (mg/dL) <sup>a</sup>	118.58 (81.42, 176.99)	130.09 (88.50, 200.89)	106.19 (75.22, 160.18)



HDL-C (mg/dL) <sup>a</sup>	50.97 (43.63, 60.61)	48.26 (40.84, 58.30)	54.05 (47.10, 62.93)
LDL-C (mg/dL)	98.60 (75.67, 122.31)	98.60 (75.87, 122.29)	98.74 (75.50, 122.41)
Hypercholesterolemia	1717 (11.1%)	910 (11.2%)	807 (11.0%)
Hypertriglyceridemia <sup>a</sup>	3494 (22.6%)	2336 (28.6%)	1158 (15.8%)
Hypoalphalipoproteinemia <sup>a</sup>	2836 (18.3%)	2064 (25.3%)	772 (10.6%)
Hyperbetalipoproteinemia	1333 (8.6%)	698 (8.6%)	635 (8.7%)

574 Abbreviations: BMI, body mass index; HDL, high-density lipoprotein cholesterol; LDL,  
575 low-density lipoprotein cholesterol; Q1, 25<sup>th</sup> percentile; Q3, 75<sup>th</sup> percentile; SD, standard  
576 deviation; TC, total cholesterol; TG, triglycerides.

577 <sup>a</sup> Statistically significant difference between men and women (p<.05).

578 **Table 2** Summary statistics and pairwise Spearman correlations of air pollutants.

Exposure	Summary statistics						Spearman correlation coefficients					
	Mean	Median	Minimum	Maximum	IQR	>WHO guideline (%) <sup>c</sup>	PM <sub>1</sub>	PM <sub>2.5</sub>	PM <sub>10</sub>	SO <sub>2</sub>	NO <sub>2</sub>	O <sub>3</sub>
PM <sub>1</sub> (µg/m <sup>3</sup> ) <sup>a</sup>	65.97	62	50	82	15	none <sup>d</sup>	1.00	0.99 <sup>e</sup>	0.73 <sup>e</sup>	0.52	0.67 <sup>e</sup>	0.47
PM <sub>2.5</sub> (µg/m <sup>3</sup> ) <sup>a</sup>	82.02	73	64	104	26	100		1.00	0.72 <sup>e</sup>	0.51	0.63 <sup>e</sup>	0.45 <sup>e</sup>
PM <sub>10</sub> (µg/m <sup>3</sup> ) <sup>b</sup>	123.06	123	93	145	19	100			1.00	0.81 <sup>e</sup>	0.65 <sup>e</sup>	0.81 <sup>e</sup>
SO <sub>2</sub> (µg/m <sup>3</sup> ) <sup>b</sup>	54.42	48	36	78	20	100				1.00	0.25	0.84 <sup>e</sup>
NO <sub>2</sub> (µg/m <sup>3</sup> ) <sup>b</sup>	35.28	33	27	45	9	18.2					1.00	0.45
O <sub>3</sub> (µg/m <sup>3</sup> ) <sup>b</sup>	49.40	50	27	71	22	0.0						1.00

579 Abbreviations: IQR, interquartile range; NO<sub>2</sub>, nitrogen dioxide; O<sub>3</sub>, ozone; PM<sub>1</sub>, particle with aerodynamic diameter ≤1.0 µm; PM<sub>2.5</sub>, particle with  
 580 aerodynamic diameter ≤2.5 µm; PM<sub>10</sub>, particle with aerodynamic diameter ≤10 µm; SD, standard deviation; SO<sub>2</sub>, sulfur dioxide.

581 <sup>a</sup>Based on values from 33 communities.

582 <sup>b</sup>Based on values from 11 districts.

583 <sup>c</sup>World Health Organization (WHO) air quality guidelines (2005).

584 <sup>d</sup>No guideline for PM<sub>1</sub>.

585 <sup>e</sup>Statistically significant correlation (p <.05).

586 **Table 3** Associations between per 10- $\mu\text{g}/\text{m}^3$  increment in air pollutants and blood lipid levels (n = 15,477).

Pollutant	%changes (95% Confidence Interval) <sup>a</sup>			
	TC	TG	HDL-C	LDL-C
PM <sub>1</sub>	1.6 (1.1, 2.0) <sup>b</sup>	2.9 (-3.3, 9.3)	-1.4 (-1.8, -0.9) <sup>b</sup>	3.2 (2.6, 3.9) <sup>b</sup>
PM <sub>2.5</sub>	1.1 (0.8, 1.4) <sup>b</sup>	1.1 (0.4, 1.8) <sup>b</sup>	-1.1 (-1.4, -0.8) <sup>b</sup>	2.9 (2.4, 3.5) <sup>b</sup>
PM <sub>10</sub>	-0.2 (-0.5, 0.1)	4.7 (3.6, 5.9) <sup>b</sup>	-0.2 (-0.7, 0.2)	-0.9 (-1.3, -0.4) <sup>b</sup>
SO <sub>2</sub>	-0.2 (-0.7, 0.1)	5.1 (3.9, 6.3) <sup>b</sup>	-0.1 (-0.6, 0.4)	-0.1 (-0.7, 0.5)
NO <sub>2</sub>	0.7 (0.0, 1.4) <sup>b</sup>	6.0 (3.5, 8.6) <sup>b</sup>	-1.6 (-2.3, -1.0) <sup>b</sup>	-0.1 (-1.2, 1.1)
O <sub>3</sub>	-1.2 (-1.6, -0.8) <sup>b</sup>	5.6 (4.5, 6.7) <sup>b</sup>	0.6 (0.2, 1.0) <sup>b</sup>	-2.7 (-3.2, -2.2) <sup>b</sup>

587 Abbreviations: HDL-C, high-density lipoprotein cholesterol; LDL-C, low-density lipoprotein cholesterol; NO<sub>2</sub>, nitrogen dioxide; O<sub>3</sub>, ozone;  
 588 PM<sub>1</sub>, particle with aerodynamic diameter  $\leq 1.0 \mu\text{m}$ ; PM<sub>2.5</sub>, particle with aerodynamic diameter  $\leq 2.5 \mu\text{m}$ ; PM<sub>10</sub>, particle with aerodynamic  
 589 diameter  $\leq 10 \mu\text{m}$ ; SO<sub>2</sub>, sulfur dioxide; TC, total cholesterol; TG, triglycerides.

590 <sup>a</sup>Adjusted for age, sex, body mass index, education, family income, smoking, alcohol drinking, exercise, diet, sugary drink intake, family  
 591 history of dyslipidemia, temperature, humidity, wind speed, season, gross domestic product, population density, residuals from regression  
 592 model of highly correlated pollutants, and study district (or community).

593 <sup>b</sup>Statistically significant association (p < .05).

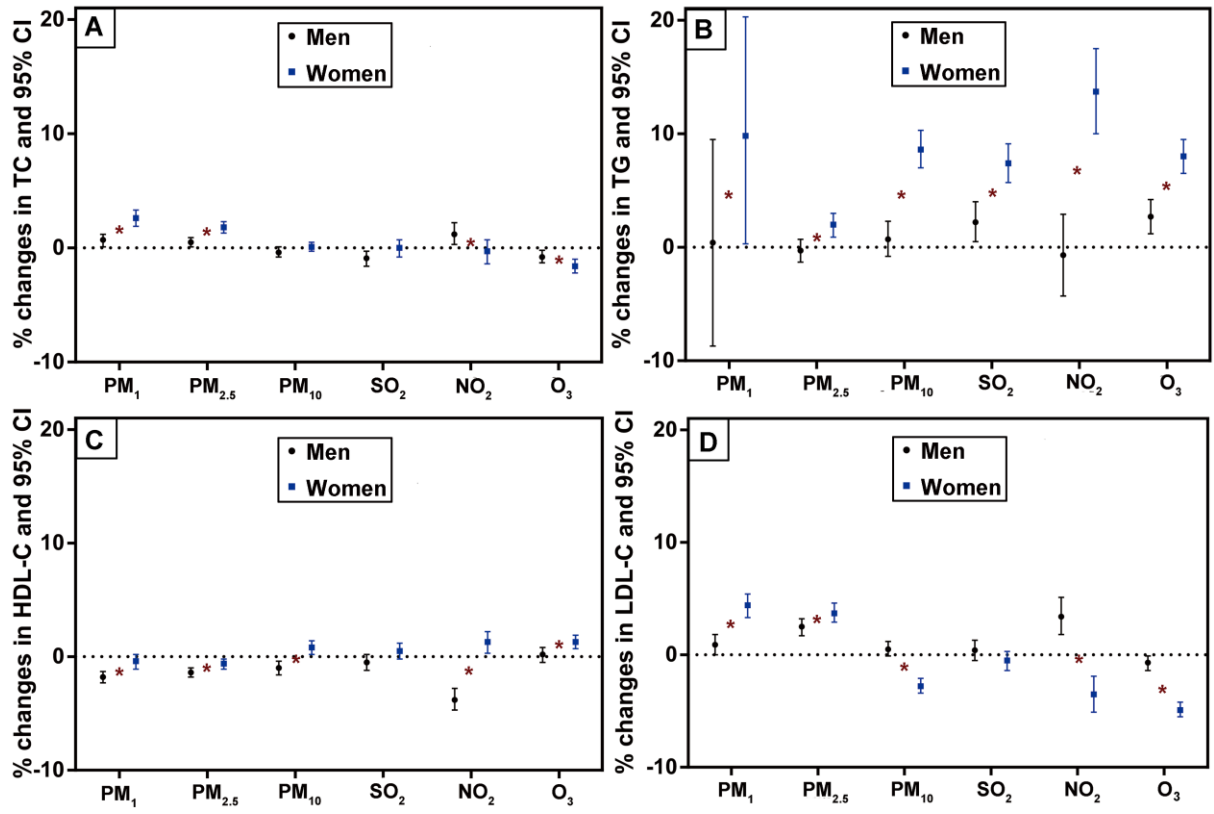
594 **Table 4** Associations between per 10- $\mu\text{g}/\text{m}^3$  increment in air pollutants and dyslipidemias (n = 15,477).

Pollutant	Odds Ratio (95% Confidence Interval) <sup>a</sup>			
	Hypercholesterolemia	Hypertriglyceridemia	Hypoalphalipoproteinemia	Hyperbetalipoproteinemia
PM <sub>1</sub>	1.26 (1.02, 1.57) <sup>b</sup>	1.03(0.91, 1.17)	1.27 (1.06, 1.52) <sup>b</sup>	1.29 (1.02, 1.64) <sup>b</sup>
PM <sub>2.5</sub>	1.18 (1.01, 1.37) <sup>b</sup>	1.07 (0.95, 1.19)	1.15 (1.02, 1.30) <sup>b</sup>	1.28 (1.05, 1.57) <sup>b</sup>
PM <sub>10</sub>	1.05 (0.92, 1.22)	1.14 (1.01, 1.29) <sup>b</sup>	1.08 (0.88, 1.32)	1.06 (0.89, 1.27)
SO <sub>2</sub>	1.11 (0.79, 1.55)	1.16 (1.00, 1.39) <sup>b</sup>	1.01 (0.79, 1.29)	0.98 (0.77, 1.25)
NO <sub>2</sub>	1.23 (1.02, 1.48) <sup>b</sup>	1.21 (0.76, 1.90)	1.27 (0.68, 2.38)	1.25 (0.74, 2.11)
O <sub>3</sub>	0.95 (0.77, 1.19)	1.17 (1.01, 1.36) <sup>b</sup>	0.97 (0.78, 1.20)	0.98 (0.84, 1.15)

595 Abbreviations: NO<sub>2</sub>, nitrogen dioxide; O<sub>3</sub>, ozone; PM<sub>1</sub>, particle with aerodynamic diameter  $\leq 1.0 \mu\text{m}$ ; PM<sub>2.5</sub>, particle with aerodynamic diameter  
 596  $\leq 2.5 \mu\text{m}$ ; PM<sub>10</sub>, particle with aerodynamic diameter  $\leq 10 \mu\text{m}$ ; SO<sub>2</sub>, sulfur dioxide.

597 <sup>a</sup>Adjusted for age, sex, body mass index, education, family income, smoking, alcohol drinking, exercise, diet, sugary drink intake, family  
 598 history of dyslipidemia, temperature, humidity, wind speed, season, gross domestic product, population density, residuals from regression  
 599 model of highly correlated pollutants, and study district (or community).

600 <sup>b</sup>Statistically significant association (p<.05).



601