

1 Short-term Effects of Fine and Coarse Particles on Deaths in Hong Kong Elderly
2 Population: an Analysis of Mortality Displacement

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7 Type of manuscript: Original research article

8 Running title: Mortality displacement by fine and coarse particles

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13 Conflicts of interest: None

14 Funding support: None

15 Acknowledgments:

16 The authors thank the Hong Kong Environmental Protection Department for providing air
17 pollution monitoring data, the Census and Statistical Department for providing mortality data,
18 and the Hong Kong Observatory for meteorological data.

19 Abbreviations:

20 DLM, distributed lag model; GAM, generalized additive model; ICD-10, international statistical
21 classification of diseases, 10th revision; IQR, inter-quartile range; PM₁₀, particulate matter with
22 an aerodynamic diameter less than 10 μm; PM_{2.5}, fine particles with an aerodynamic diameter
23 less than 2.5 μm; PM_c, coarse particles with an aerodynamic diameter between 2.5 and 10 μm.

24 Abstract

25 Background: While numerous studies worldwide have evaluated the short-term associations of
26 fine and coarse particulate matter (PM) air pollution with mortality and morbidity, these
27 studies may be susceptible to short-term harvesting effect. We aimed to investigate the short-
28 term association between mortality and PM with aerodynamic diameter less than 2.5 μm
29 ($\text{PM}_{2.5}$) and those between 2.5 and 10 μm (PMc) within a month prior to death, and assess the
30 mortality displacement by $\text{PM}_{2.5}$ and PMc among elderly population in Hong Kong.

31 Methods: We obtained air pollution data from January 2011 to December 2015 from
32 Environmental Protection Department, and daily cause-specific mortality data from Census and
33 Statistical Department of Hong Kong. We performed generalized additive distributed lag model
34 to examine the acute, delayed and long-lasting effects of $\text{PM}_{2.5}$ and PMc within one month on
35 mortality.

36 Results: We observed a statistically significant association of $\text{PM}_{2.5}$ and PMc exposure over lags
37 0-6 days with all natural mortality and cardio-respiratory mortality. The overall cumulative
38 effect of $\text{PM}_{2.5}$ over 0-30 lag days was 3.44% (95% CI: 0.30-6.67%) increase in all natural
39 mortality and 6.90% (95% CI: 0.58-13.61%) increase of circulatory mortality, which suggested
40 the absence of mortality displacement by $\text{PM}_{2.5}$. On the other hand, no significant cumulative
41 association with mortality was found for PMc over 0-30 lag exposure window, and thus
42 mortality displacement by PMc cannot be ruled out. Findings remained robust in various
43 sensitivity analyses.

44 Conclusions: We found adverse effect of both $\text{PM}_{2.5}$ and PMc exposure within one week prior
45 to death. While there was no evidence of mortality displacement in the association of $\text{PM}_{2.5}$
46 exposure over one month prior with all natural and circulatory mortality, mortality
47 displacement by PMc cannot be ruled out. $\text{PM}_{2.5}$ may contribute more to the longer term effect
48 of particulate matter than PMc .

49 Key Words: Coarse Particulate Matter; Fine Particulate Matter; Generalized Additive
50 Distributed Lag Model; Mortality Displacement; Time Series Study

51 Capsule: This time series study with analysis of mortality displacement demonstrated the acute
52 effects of both fine and coarse PM on mortality within one week and no evidence of mortality
53 displacement by PM_{2.5} within one month prior to death.

54 Introduction

55 The associations of fine and coarse particulate matter (PM) air pollution with mortality and
56 morbidity have been widely discussed (Adar et al., 2014; Brunekreef and Forsberg, 2005), with
57 consistent evidence of an acute health effect of fine particles with aerodynamic diameter ≤ 2.5
58 μm (PM_{2.5}) (Englert, 2004), and supporting evidence of an effect of coarse particles with
59 aerodynamic diameter between 2.5-10 μm (PM_c), especially with respiratory diseases (Adar et
60 al., 2014; Qiu et al., 2012). While majority of the epidemiological studies on the health effects
61 of PM_{2.5} and PM_c focused on short exposure period, i.e., from the day of disease onset/death
62 to previous few days (Meister et al., 2012; Qiu et al., 2012; Samoli et al., 2013; Zanobetti and
63 Schwartz, 2009), these studies may be susceptible to potential short-term harvesting effect
64 (Schwartz, 2000a). Harvesting effect, also known as mortality displacement, of air pollution-
65 related deaths is a phenomenon where air pollution principally affects frail population by
66 advancing their deaths by a number of days or weeks because of their already poor health
67 conditions, and the initial increase in mortality rate is then followed by a period with a lower-
68 than-expected mortality rate (Schwartz, 2001; Zeger et al., 1999). The presence of harvesting
69 effect could limit the public health significance of air pollution.

70 A combination of *generalized additive model* and *distributed lag model* has been suggested to
71 model the relationship between lagged exposure of multiple days (e.g., lag 0 to 30 days), and
72 subsequently to quantify the mortality displacement or harvesting effect effectively in air
73 pollution epidemiological studies (Schwartz, 2000a; Zanobetti et al., 2000). However, there
74 remains limited research on the harvesting effect and distributed lag effects of PM. In 2000,
75 Schwartz examined various distributed lag of PM_{2.5} on mortality using seasonal-trend
76 decomposition algorithm, and found evidence of harvesting effects on different time scales
77 (Schwartz, 2000b). Costa et al. showed evidence of mortality displacement within 30 days for
78 nonaccidental and circulatory deaths using distributed lag model in elderly residents of São
79 Paulo (Costa et al., 2017). To our knowledge, no studies have assessed the harvesting effect and
80 distributed lag effects of PM_{2.5} and PM_c in the same setting.

81 Hong Kong is an ideal place to study the health effects of short-term exposure to PM_{2.5} and
82 PM_c, because of its readily available measures of hourly PM₁₀ and PM_{2.5} concentrations

83 monitored simultaneously in every monitoring station dispersed across the territory since
84 January 2011. Our previous studies have demonstrated acute effects of both PM_{2.5} and PMc
85 within one week on cardio-pulmonary diseases in Hong Kong (Qiu et al., 2014, 2013, 2012). The
86 current study built upon this observation by quantifying mortality displacement and examining
87 the short-term association between cause-specific mortality and PM_{2.5} and PMc exposure in a
88 month prior to death, using the package 'dlnm' developed within the statistical environment R
89 (Gasparrini et al., 2010). Given that the aging population in Hong Kong is increasing [from 12%
90 in 2006 to 16% in 2016 (<http://www.censtatd.gov.hk>)], and because they are most vulnerable
91 to air pollution, we studied the mortality displacement in Hong Kong elderly population aged 65
92 or above.

93

94 Materials and Methods

95 Data collection

96 We obtained pairwise hourly measures of PM_{2.5} and PM₁₀ concentrations collected between
97 January 1, 2011 to December 31, 2015 at 14 general air quality monitoring stations maintained
98 by the Hong Kong Environmental Protection Department (EPD) (HKEPD, 2016). Missing data
99 accounted for only 3.1% and 10.7% of PM₁₀ and PM_{2.5} measurements, respectively, and thus
100 data were not imputed. Hourly PMc concentrations were calculated by subtracting PM_{2.5} from
101 PM₁₀ measurements for each station. We then computed daily 24-hr mean concentrations of
102 PM_{2.5} and PMc if at least 18 of 24 hourly measurements were available for each monitor. Air
103 pollution measurements from one general station on a remote island and three roadside
104 stations were excluded, and the final analysis included daily measurements of PM_{2.5} and PMc
105 from 10 general monitoring stations that represent the citywide background air pollution level
106 and general population's daily exposure (Qiu et al., 2014). Meteorological data of daily mean
107 temperature and relative humidity were collected from Hong Kong Observatory for the same
108 study period.

109 Mortality data among elderly Hong Kong residents aged 65 or above between 2011 and 2015
110 were obtained from Hong Kong Census and Statistical Department. The causes of death were

111 identified according to the WHO *International Statistical Classification of Diseases, 10th Revision*
 112 (ICD-10). Daily counts of mortality from all natural causes (ICD-10: A00-R99), circulatory
 113 diseases (ICD-10: I00-I99) and respiratory diseases (ICD-10: J00-J99) were computed and
 114 subsequently linked to air pollution and meteorological data. Ethics approval and consent from
 115 individual subjects were not required, as no individualized data but aggregated data were used
 116 in this study.

117

118 Statistical modelling

119 We used time series Poisson model to examine the association between PM in different size
 120 fractions and daily mortality. Generalized additive regression model (GAM) integrated with
 121 distributed lag model (DLM) were used to investigate the potential mortality displacement in
 122 the association (Zanobetti et al., 2000). We used smoothing spline function, $s(\cdot)$, to filter out
 123 long-term trend and seasonality in time series of daily mortality as well as daily mean
 124 temperature and relative humidity (Peng et al., 2006; Schwartz et al., 1996). Based on previous
 125 studies we followed a priori model specifications and the degree of freedom (df) for the time
 126 trend and other time-varying covariates, in order to reduce the problems coming from multiple
 127 testing and model selection strategies (Peng et al., 2006; Qiu et al., 2012). Time trend with a df
 128 of 7 per year, temperature of the current day ($Temp_0$) and the mean of previous 3 days ($Temp_{1-3}$)
 129 with a df of 6, and relative humidity of the current day ($Humid_0$) with a df of 3 were used. The
 130 day of the week (DOW) and public holidays ($Holiday$) were included as dummy variables to the
 131 model (Qiu et al., 2012).

132 Briefly, a core model was set up to remove the long-term trend, seasonality, with adjustment
 133 for time varying covariates as follows:

$$134 \log[E(Y)] = a + s(t, df=7/year \times 5years) + s(Temp_0, df=6) + s(Temp_{1-3}, df=6) + s(Humid_0, df=3)$$

$$135 \quad + \beta_{1,i} DOW + \beta_2 Holiday \dots\dots\dots(1)$$

136 Here $E(Y)$ is the expected daily counts of mortality on day t ; $s(\cdot)$ is the smoothing spline function
 137 for nonlinear covariates. We observed no discernible patterns and autocorrelation in the model

138 residuals assessed by residual plot and partial autocorrelation function (PACF) for all three
139 mortality outcomes (Costa et al., 2017). The standardized deviance residuals shown in the Q-Q
140 plot were also normal (Figure 1), which suggested that all unmeasured time-varying
141 confounding in the daily variations of mortality series had been controlled for. Once the model
142 was correctly specified, the terms of PM_{2.5} or PMc were included into the model to estimate
143 their single-pollutant association with daily cause-specific mortality.

144 Distributed lag model (DLM), which was integrated in the GAM as *cross-basis* function to
145 account for the potential distributed and lagged effect of pollution on mortality flexibly, was
146 used to estimate the short-term health effects associated with PM_{2.5} or PMc exposure in the 30-
147 days prior to mortality (Gasparrini et al., 2010). In the primary analysis, second-degree
148 (quadratic) polynomials was used to constrain the smooth shape of the effects of distributed
149 lags ≤ 30 days for daily mortality (Costa et al., 2017). We calculated the cumulative effect of
150 PM_{2.5} or PMc distributed over 0-30 lag days to estimate the overall effect lasting for one month,
151 as well as the cumulative effects of PM_{2.5} or PMc over the different lag periods: 0-6 days, 7-13
152 days, and 14-30 days, representing the acute, delayed and long-lasting effects, respectively (Qiu
153 et al., 2016). Single-lag effects of PM_{2.5} and PMc over 30 days exposure on daily mortality were
154 shown, as well as the cumulative effects of PM_{2.5} and PMc for the investigation of mortality
155 displacement (Costa et al., 2017; Gasparrini, 2011). As described in detail by Schwartz and
156 Zanobetti, the cumulative effects of PM on daily mortality should reduce to zero, and the
157 confidence intervals for the cumulative risks should include zero in the presence of mortality
158 displacement (Schwartz, 2000a; Zanobetti et al., 2000).

159 We conducted sensitivity analysis by repeating the regression analysis using an unconstrained
160 DLM, which has been shown to give unbiased estimate of the cumulative effect even though it
161 is more noisy than their constrained counterpart to provide information about the shape of the
162 associations along the lags (Zanobetti et al., 2003, 2002). We also constructed a cubic
163 polynomial DLM to test the robustness of our effect estimates against different degrees of
164 polynomial. Moreover, exposure windows of 0-20 and 0-40 lag days were examined,
165 respectively, to compare with lag period over 0-30 days in the main analysis. Two-pollutant

166 model with PM_{2.5} and PM_c over 0-30 lag days included simultaneously in the model was
167 constructed to assess the independent effect of each pollutant on mortality.
168 The effect estimates were presented as the percentage changes of daily deaths for an
169 interquartile range (IQR) increase of PM_{2.5} or PM_c, and their corresponding 95% confidence
170 intervals (CI). All analyses were conducted in the statistical environment R 3.3.3 while loading
171 'mgcv' and 'dlnm' packages (R Development Core Team, 2017: <http://www.r-project.org>).

172

173 Results

174 Between January 1, 2011 and December 31, 2015, a total of 168,541 all natural deaths in Hong
175 Kong elderly population (age > 65 years) were reported; among which 42,264 deaths were
176 attributed to circulatory diseases and 44,810 deaths from respiratory diseases. On average,
177 there were 92, 23 and 24 deaths from all natural causes, circulatory diseases and respiratory
178 diseases, respectively, per day. The daily 24-hour mean concentrations were 29.1 µg/m³ for
179 PM_{2.5} and 14.6 µg/m³ for PM_c, while the corresponding IQRs were 24.9 and 10.0 µg/m³,
180 respectively. Pearson correlation coefficient between PM_{2.5} and PM_c was 0.71. The daily mean
181 temperature was 23.5°C and the relative humidity was 78.3% (Table 1).

182 Table 2 shows the cumulative effects of PM_{2.5} and PM_c on daily mortality over the different lag
183 periods in single-pollutant models. We observed a statistically significant association of PM_{2.5}
184 exposure over lags 0-6 days with all natural mortality (3.23%, 95% CI: 1.85-4.63%), circulatory
185 mortality (4.81%, 95% CI: 2.06-7.63%), as well as respiratory mortality (3.74%, 95% CI: 0.97-
186 6.57%). The associations between PM_{2.5} and for all natural and circulatory mortality persisted
187 for PM_{2.5} exposure over 7-13 days prior to death, though the magnitude of the effects
188 attenuated. No significant associations with respiratory mortality were observed for PM_{2.5}
189 exposure after lag 6 days, and no associations with all natural and circulatory mortality after lag
190 14 days (i.e., lag 14-30 days). The overall cumulative effect of PM_{2.5} over 0-30 lag days was
191 3.44% (95% CI: 0.30-6.67%) increase in all natural mortality, and 6.90% (95% CI: 0.58-13.61%)
192 increase in circulatory mortality. Since the effect estimates were not zero and the confidence

193 intervals didn't include zero, no mortality displacement for all natural and circulatory deaths by
194 $PM_{2.5}$ was observed over 0-30 lag exposure window.

195 Similar to $PM_{2.5}$, increment in PMc exposure over 0-6 lag days was also significantly associated
196 with increase of 2.90% (95% CI: 1.68-4.14%), 2.77% (0.34-5.25%) and 4.19% (1.76-6.69%) for all
197 natural deaths, circulatory deaths and respiratory deaths, respectively. However, the effects of
198 PMc attenuated and rendered insignificant for exposure over 7-13 days prior to death,
199 suggesting a less lasting effect of PMc on mortality than that of $PM_{2.5}$. Since the confidence
200 intervals of the cumulative effects of PMc over 0-30 lag days included zero, mortality
201 displacement by PMc within one month cannot be ruled out (Table 2). Figures 2 and 3 showing
202 single-lag effects and cumulative effects of $PM_{2.5}$ and PMc, respectively, also provided
203 consistent findings of acute effects of both $PM_{2.5}$ and PMc on mortality within one week, and
204 the absence of harvesting effects in the association between $PM_{2.5}$ and all natural mortality and
205 circulatory mortality over 30-days lag window.

206 Findings from sensitivity analyses using unconstrained DLM or cubic polynomial DLM, or longer
207 exposure windows were similar to those in the primary analysis (Table 3). Two-pollutant
208 models with $PM_{2.5}$ and PMc adjusted simultaneously showed that the association of $PM_{2.5}$ with
209 mortality persisted whereas that of PMc rendered insignificant upon controlling for $PM_{2.5}$
210 (Table 3).

211

212 Discussion

213 This is the first study that investigated the distributed lag effects of $PM_{2.5}$ and PMc exposure in
214 30-days prior to death on cause-specific mortality, and assessed whether mortality
215 displacement existed in the associations. We observed both $PM_{2.5}$ and PMc to be associated
216 with significant increase in all natural, cardio-respiratory mortality in lag 0-6 days, which are
217 consistent with findings from existing literature of an adverse association between short-term
218 exposure of $PM_{2.5}$ and PMc and mortality (Lee et al., 2015; López-Villarrubia et al., 2012; Malig
219 and Ostro, 2009; Meister et al., 2012; Perez et al., 2009; Yorifuji et al., 2016; Zanobetti and
220 Schwartz, 2009). However, only $PM_{2.5}$ was found to be significantly linked to all natural and

221 circulatory mortality in lag 0-30 days, suggesting the absence of mortality displacement by
222 PM_{2.5}. Yet, mortality displacement by PM_{2.5} in the PM_{2.5}-respiratory association or by PMc
223 cannot be ruled out.

224 Common plausible mechanisms for the association between PM and cardio-respiratory health
225 include the oxidative stress and inflammation pathways. Fine particles could trigger redox-
226 sensitive pathways via catalytic generation of reactive oxygen species, leading to inflammation
227 and cell death (Lodovici and Bigagli, 2011). Studies with healthy volunteers demonstrated that
228 coarse concentrated PM exposure may induce pulmonary inflammation, decrease blood tissue
229 plasminogen activator (Graff et al., 2009), influence systemic biomarkers (Graff et al., 2009; Liu
230 et al., 2015), as well as trigger the hemodynamic alternations through elevating systolic blood
231 pressure and heart rate (Morishita et al., 2015). High level of PMc was associated with
232 reduction in heart rate variability among older subjects (Graff et al., 2009; Lipsett et al., 2006).
233 An animal study also suggested that the chemical compositions of PM_{2.5} and PMc may be
234 responsible for inflammation and lung tissue damage (Happo et al., 2010).

235 Harvesting effect or mortality displacement principally exists when the mortality rate of
236 susceptible population decreases following an increase in air pollution-associated mortality
237 rate, resulting in an inverse effect of air pollution in longer exposure lags (Schwartz, 2000a;
238 Zanobetti et al., 2000). We extended our exposure window to 0-30 lag days, and observed the
239 statistically significant cumulative effects of PM_{2.5} on all natural and circulatory mortality, which
240 suggests the absence of mortality displacement over this window. Meanwhile, we cannot rule
241 out possible harvesting effect of PM_{2.5} on respiratory mortality, as well as harvesting effect of
242 PMc on all natural and cardiorespiratory deaths within one month. A limited number of studies
243 have investigated potential short-term mortality displacement in the association of air pollution
244 and mortality, but these studies only assessed the harvesting for PM₁₀ (Costa et al., 2017;
245 Goodman et al., 2004; Schwartz, 2001; Zanobetti et al., 2003, 2002), PM_{2.5} or black smoke
246 (Goodman et al., 2004; Schwartz, 2000b). Our findings are similar to those from Schwartz, 2000,
247 and further contributed to the existing literature by showing potential evidence of mortality
248 displacement by PMc. While direct comparison is not possible, our results may be different
249 from those in a recent study conducted in Brazil, in which Costa et al. provided evidence of

250 mortality displacement for PM₁₀ within 30 days for all natural and circulatory deaths but not for
251 respiratory and cancer deaths (Costa et al., 2017). The different population characteristics,
252 weather conditions and ambient pollution profiles, together with different disease spectrum in
253 San Paulo and Hong Kong may contribute to the diverse results. Differential harvesting effects
254 of fine and coarse particles examined in other places of the world are required to better
255 understand sources of heterogeneity.

256 Findings from our study also inferred that PM_{2.5} may contribute more in the longer-term effect
257 of PM₁₀ exposure than the PMc. Consistent with our study, the Nurses' Health Study also
258 reported evidence of a stronger chronic PM_{2.5} effect on risk of all-cause and cardiovascular
259 mortality than that of PMc (Puett et al., 2009). Furthermore, two-pollutant models adjusting for
260 PM_{2.5} and PMc simultaneously also supported less independent effect of coarse fraction.
261 Compared with PMc, PM_{2.5} has higher number concentration, greater surface area, and better
262 lung deposition (Dockery, 2009). PM_{2.5} has a greater probability of penetrating deeply into the
263 small airways and the alveoli of the lung (Dominici et al., 2006), whereas PMc is more likely to
264 affect the upper and larger airways (Host et al., 2008). Since particles in the alveolar region are
265 removed at a slower rate than those in the conducting airways, this may explain the greater
266 toxic effect of PM_{2.5} (Englert, 2004).

267 Our study had several strengths. We are the first study to assess mortality displacement by
268 PM_{2.5} and PMc in the same setting, as previous research focused only on PM₁₀ or black smoke
269 probably due to data scarcity. Hong Kong is an ideal Asian city for studying the health effect of
270 air pollution, with its greater variability in air pollution than that in most of western countries.
271 Also, the availability of high quality of hourly PM₁₀ and PM_{2.5} measurements from a descent
272 network of air monitoring stations across the entire Hong Kong territory provided more
273 representative assessment for population exposure than those used data monitored from only
274 one fixed-site station (Kan et al., 2007) or every sixth day sampling window (Peng et al., 2008).

275 Some limitations of the study should be considered. Firstly, we calculated PMc concentrations
276 by subtracting PM_{2.5} from PM₁₀ measurements, thus the PMc estimation may be susceptible to
277 double measurement errors. Pollutant with large measurement error may lose statistical
278 significance in co-pollutant model, which may explain the null findings of PMc upon adjusting

279 for PM_{2.5}. Secondly, exposure misclassification cannot be eliminated in our time series study
280 design. While ambient PM_{2.5} concentrations measured from local monitoring stations might be
281 adequate surrogate for the total personal exposures (Schwartz et al., 2007), the
282 representativeness of ambient PMc concentrations is less certain. Moreover, we did not
283 estimate season-specific effects as stratification could lead to more missing data in the 30-day
284 PM exposure assessment, thereby increasing the variability and bias of the effect estimates.
285 However, seasonal analysis using a shorter exposure lag (i.e., lag0-6 days) for PM_{2.5} and PMc
286 showed stronger association with mortality for the cold season as compared to those in the
287 warm season (data not shown); this finding is generally consistent with previous literature
288 (Meister et al., 2012; Zanobetti and Schwartz, 2009). Finally, residual confounding remains
289 possible as we did not adjust for potential confounding from gaseous pollutants in
290 consideration of collinearity, unstable estimates, and to avoid the problems of model
291 misconvergence.

292

293 Conclusions

294 In conclusion, we found adverse effect of both PM_{2.5} and PMc exposure within one week prior
295 to death. While there was no evidence of mortality displacement in the association between all
296 natural and circulatory mortality and PM_{2.5} exposure over one month prior to death, mortality
297 displacement by PMc cannot be ruled out. PM_{2.5} may contribute more to the longer term effect
298 of PM air pollution and impose stronger public health impact than PMc.

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420 Table-1. Descriptive Statistics of Daily Mortality Counts in the Elders, Air Pollution
 421 Concentrations and Weather Conditions in Hong Kong, 2011-2015 (1826 days)

	Mean	SD	Min	25th	50th	75th	Max
Daily Mortality Counts							
All Natural Deaths	92.3	15.8	48.0	81.2	90.0	101.1	150.0
Circulatory Deaths	23.1	6.5	8.0	19.0	22.0	27.0	52.0
Respiratory Deaths	24.5	7.1	8.0	19.0	24.0	29.0	53.0
Pollution concentration							
PM ₁₀ (µg/m ³)	43.7	23.8	7.6	24.0	38.9	57.8	157.4
PM _{2.5} (µg/m ³)	29.1	17.3	4.9	14.6	25.9	39.5	115.7
PMc (µg/m ³)	14.6	8.2	2.4	8.7	12.6	18.7	108.9
Weather conditions							
Temperature (°C)	23.5	5.3	8.4	19.0	24.8	28.2	32.4
Relative humidity (%)	78.3	10.3	29.0	74.0	79.0	85.0	99.0

422 Abbreviations: SD-standard deviation; Min-minimum; Max-maximum; PM₁₀, particulate matter with
 423 aerodynamic diameter less than 10 µm; PM_{2.5}, particulate matter with aerodynamic diameter less than
 424 2.5 µm; PMc, particulate matter with aerodynamic diameter between 2.5 and 10 µm; NO₂, nitrogen
 425 dioxide; SO₂, sulfur dioxide; O₃, ozone.

426 Table-2 Cumulative effects (% change with 95% CIs) in Daily Mortality in the Elderly
 427 Associated with an IQR Increase of PM over Different Lag Periods in Hong Kong, 2011-2015*

	All Natural Death	Circulatory Death	Respiratory Death
PM_{2.5}			
Lags 0-6	3.23 (1.85, 4.63)	4.81 (2.06, 7.63)	3.74 (0.97, 6.57)
Lags 7-13	0.87 (0.04, 1.70)	1.79 (0.15, 3.46)	1.04 (-0.61, 2.71)
Lags 14-30	-0.67 (-2.65, 1.36)	0.19 (-3.71, 4.26)	0.04 (-3.94, 4.18)
Lags 0-30	3.44 (0.30, 6.67)	6.90 (0.58, 13.61)	4.85 (-1.44, 11.54)
PM_c			
Lags 0-6	2.90 (1.68, 4.14)	2.77 (0.34, 5.25)	4.19 (1.76, 6.69)
Lags 7-13	0.44 (-0.27, 1.15)	1.10 (-0.30, 2.52)	-0.18 (-1.57, 1.23)
Lags 14-30	-1.84 (-3.54, -0.11)	0.14 (-3.25, 3.66)	-2.32 (-5.64, 1.12)
Lags 0-30	1.45 (-1.30, 4.29)	4.05 (-1.48, 9.89)	1.59 (-3.81, 7.30)

428 *: Interquartile ranges (IQRs) for PM_{2.5}, and PM_c are 24.9, and 10.0 µg/m³, respectively. Effects
 429 are estimated from Poisson generalized additive distributed lag model, constrained with a
 430 second-degree (quadratic) polynomial, while adjusting for time trend and seasonality, weather
 431 conditionals, day of week and public holidays. The highest 0.5% and the lowest 0.5% extreme
 432 values of PM_c have been excluded from the analysis. Statistically significant effect estimates are
 433 in bold.

434 Table-3 Sensitivity analyses showing the cumulative effects (% change with 95% CIs) in Daily
 435 Mortality in the Elderly per IQR Increase of PM in Hong Kong, 2011-2015*

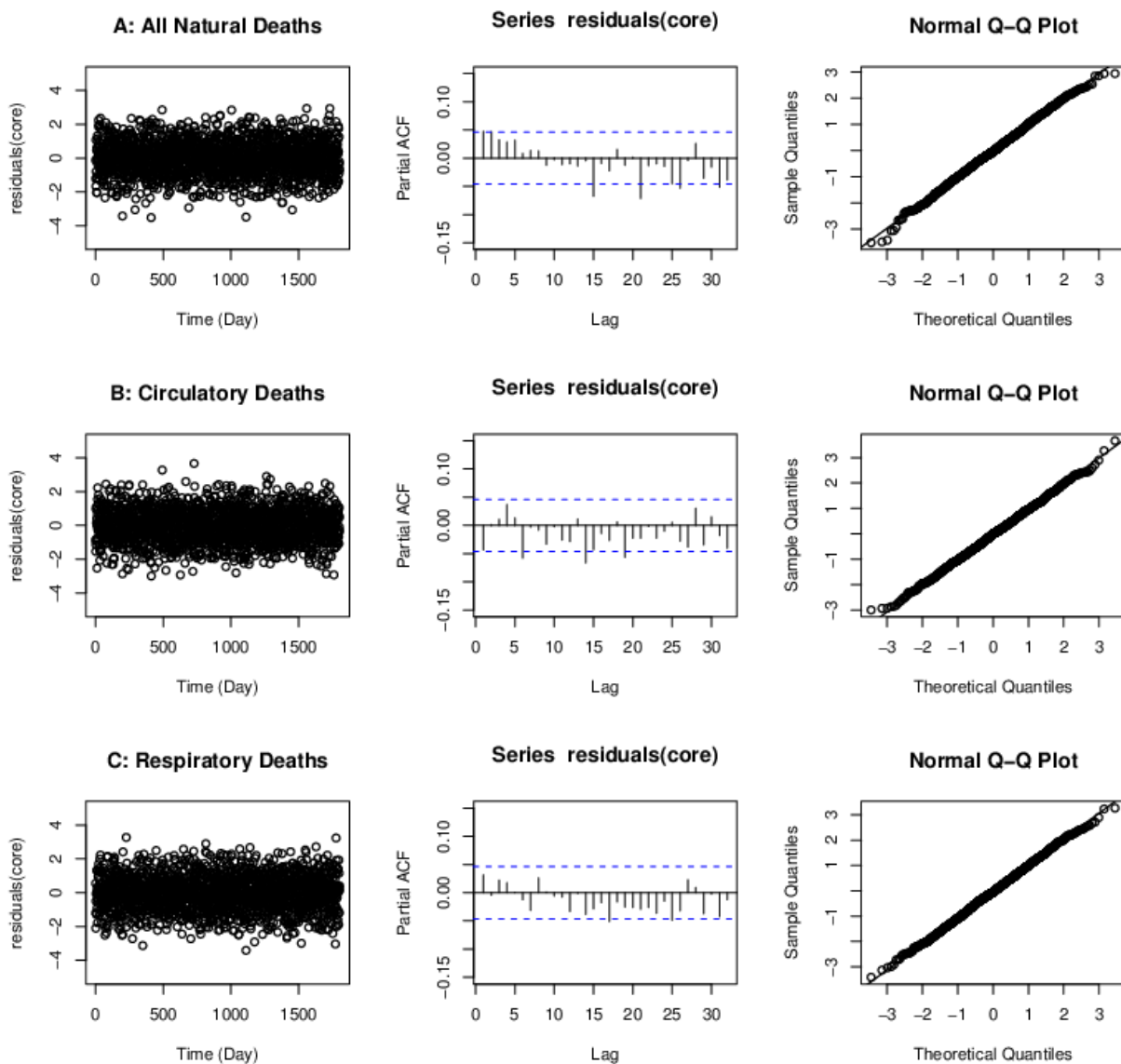
	All Natural Death	Circulatory Death	Respiratory Death
Cubic polynomial DLM over 0-30 lag days			
PM _{2.5}	3.43 (0.30, 6.67)	6.88 (0.56, 13.60)	4.85 (-1.44, 11.54)
PMc	1.42 (-1.34, 4.26)	3.97 (-1.56, 9.81)	1.62 (-3.78, 7.33)
Unconstrained DLM over 0-30 lag days			
PM _{2.5}	3.31 (0.11, 6.62)	6.44 (-0.02, 13.31)	4.37 (-2.06, 11.23)
PMc	1.20 (-1.65, 4.13)	2.73 (-2.92, 8.71)	1.73 (-3.88, 7.67)
Quadratic polynomial DLM over 0-20 lag days			
PM _{2.5}	3.14 (0.72, 5.62)	6.52 (1.65, 11.62)	4.01 (-0.83, 9.09)
PMc	1.62 (-0.5, 3.78)	3.63 (-0.58, 8.03)	0.19 (-3.91, 4.46)
Quadratic polynomial DLM over 0-40 lag days			
PM _{2.5}	2.61 (-1.70, 7.12)	8.93 (0.05, 18.61)	-1.26 (-9.45, 7.66)
PMc	2.29 (-1.46, 6.19)	7.08 (-0.57, 15.33)	1.91 (-5.43, 9.83)
Quadratic polynomial DLM over 0-30 lag days in two-pollutant model			
PM _{2.5}	5.60 (0.69, 10.75)	8.09 (-1.59, 18.72)	9.36 (-0.52, 20.23)
PMc	-2.29 (-6.39, 1.98)	-1.30 (-9.26, 7.35)	-4.59 (-12.29, 3.80)

436 *: Interquartile ranges (IQRs) for PM_{2.5} and PMc are 24.9 and 10.0 µg/m³, respectively. Effects
 437 are estimated from Poisson generalized additive distributed lag model, while adjusting for time
 438 trend and seasonality, weather conditionals, day of week and public holidays. The highest 0.5%
 439 and the lowest 0.5% extreme values of PMc have been excluded from the analysis. Statistically
 440 significant effect estimates are in bold.

441 Figure legends:

442 Figure-1 Diagnostic plots of the residuals of the core model for A: All Natural Deaths; B:

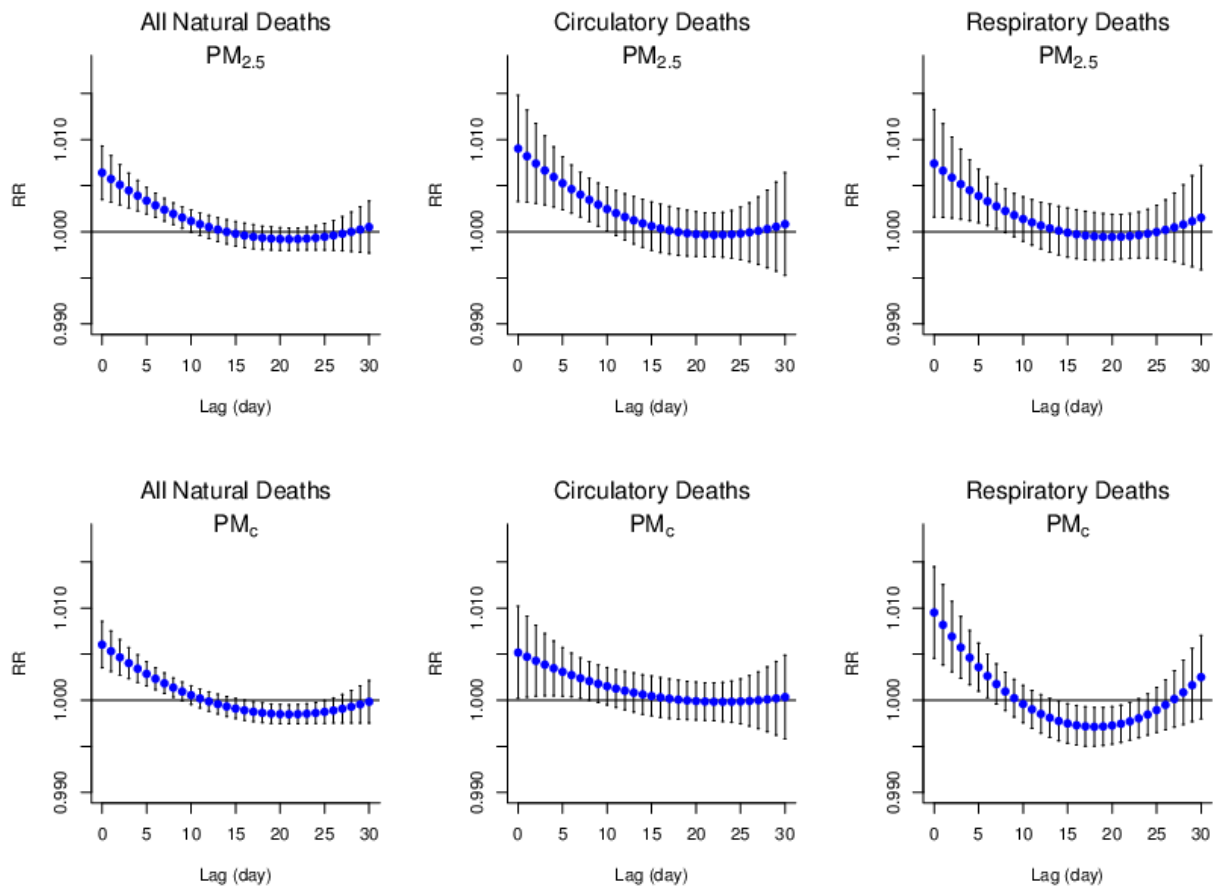
443 Circulatory Deaths; C: Respiratory Deaths



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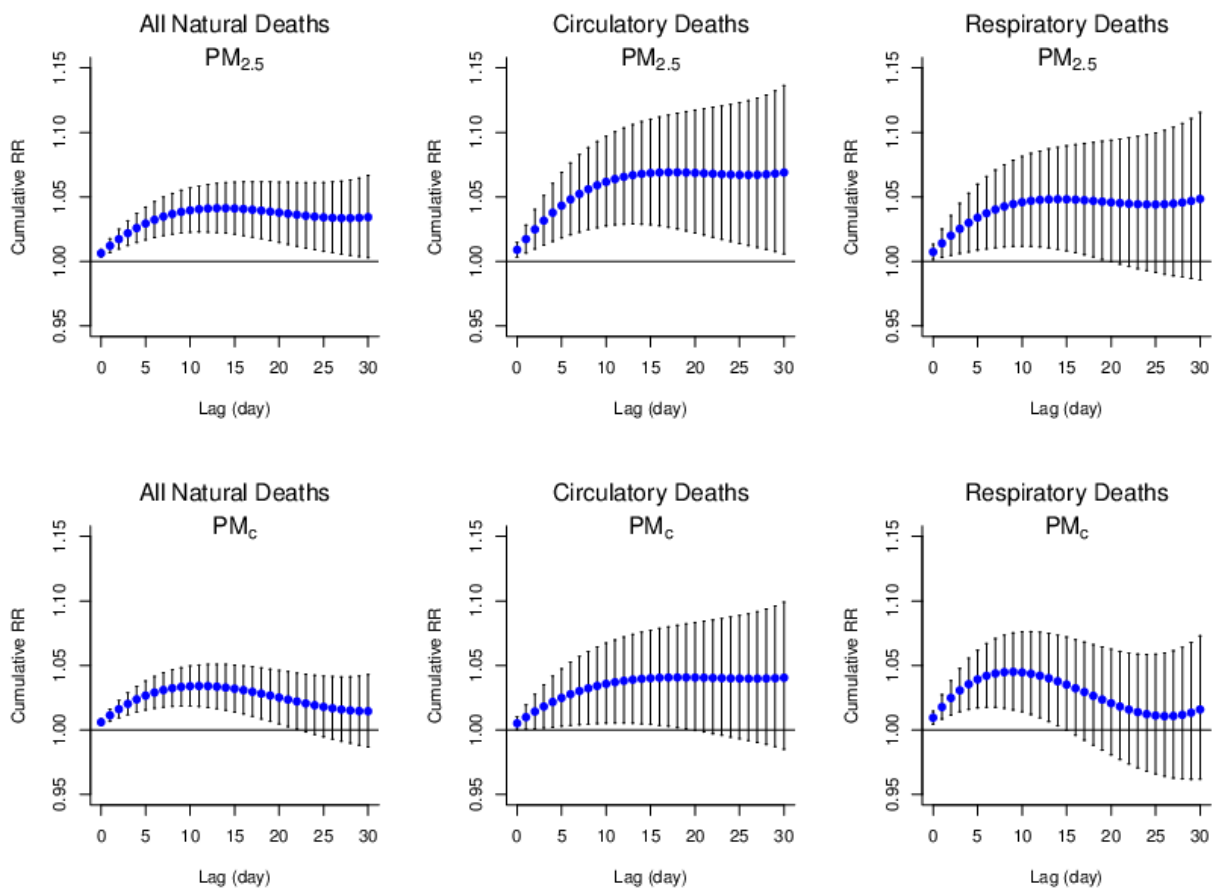
446 Figure-2 Single-lag effects of PM_{2.5} and PM_c per IQR increase over 0-30 lag days on daily
447 mortality of Hong Kong Elderly population, 2011-2015



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450 Figure-3 Cumulative effects of PM_{2.5} and PM_c per IQR increase over 0-30 lag days on daily
451 mortality of Hong Kong Elderly population, 2011-2015



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