Chronic Exposure to Fine Particles and Mortality: An Extended Follow-up of the Harvard Six Cities Study from 1974 to 2009

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BACKGROUND: Epidemiologic studies have reported associations between fine particles (aerodynamic diameter ≤ 2.5 μm; $PM_{2.5}$) and mortality. However, concerns have been raised regarding the sensitivity of the results to model specifications, lower exposures, and averaging time.

OBJECTIVE: We addressed these issues using 11 additional years of follow-up of the Harvard Six Cities study, incorporating recent lower exposures.

METHODS: We replicated the previously applied Cox regression, and examined different time lags, the shape of the concentration–response relationship using penalized splines, and changes in the slope of the relation over time. We then conducted Poisson survival analysis with time-varying effects for smoking, sex, and education.

RESULTS: Since 2001, average $PM_{2.5}$ levels, for all six cities, were < 18 μg/m³. Each increase in $PM_{2.5}$ (10 μg/m³) was associated with an adjusted increased risk of all-cause mortality ($PM_{2.5}$ average on previous year) of 14% [95% confidence interval (CI): 7, 22], and with 26% (95% CI: 14, 40) and 37% (95% CI: 7, 75) increases in cardiovascular and lung-cancer mortality ($PM_{2.5}$ average of three previous years), respectively. The concentration–response relationship was linear down to $PM_{2.5}$ concentrations of 8 μg/m³. Mortality rate ratios for $PM_{2.5}$ fluctuated over time, but without clear trends despite a substantial drop in the sulfate fraction. Poisson models produced similar results.

CONCLUSIONS: These results suggest that further public policy efforts that reduce fine particulate matter air pollution are likely to have continuing public health benefits.

KEY WORDS: air pollution, cohort studies, concentration–response, follow-up studies, lag, lung cancer, mortality, particles, $PM_{2.5}$, threshold. *Environ Health Perspect* 120:965–970 (2012). http://dx.doi.org/10.1289/ehp.1104660 [Online 28 March 2012]

All-cause, cardiopulmonary, cardiovascular, and lung-cancer mortality have been associated with chronic air pollution exposure in prospective studies that controlled for individual covariates (Abbey et al. 1999; Beelen et al. 2008b; Beeson et al. 1998; Cao et al. 2011; Dockery et al. 1993; Eftim et al. 2008; Filleul et al. 2005; Gehring et al. 2006; Katanoda et al. 2011; Laden et al. 2006; Miller et al. 2007; Nafstad et al. 2004; Ostro et al. 2010; Pope et al. 2002; Puett et al. 2009; Yorifuji et al. 2011). The studies that specifically considered lung-cancer mortality associations with fine particles (aerodynamic diameter < 2.5 μm; $PM_{2.5}$), all found positive associations (Beelen et al. 2008b; Dockery et al. 1993; Laden et al. 2006; McDonnell et al. 2000), although this association was only statistically significant ($p < 0.05$) in the American Cancer Society study (ACS) (Pope et al. 2002; Turner et al. 2011).

Although compelling evidence supports the harmful effects of $PM_{2.5}$ on longevity, concerns have been raised regarding the sensitivity of the results to model specifications. In particular, Moolgavkar (2005, 2007) suggested that covariates may not be proportional and hence were not controlled for properly in proportional hazards models; that the concentration–response relation may not be linear; and that there are few observations at levels as low as or below the

current World Health Organization and U.S. Environmental Protection Agency (EPA) air quality standards. In addition, the relative toxicity of particle elements is still controversial, and most of the recent reduction in $PM_{2.5}$ concentrations in the United States has come from sulfate control. Hence it is of interest whether the concentration–response curve has changed over time as particle composition has changed. Health impact assessments in the United States assume that health benefits of reducing particles are only fully realized after 20 years (U.S. EPA 2010), so examination of the lag between exposure and mortality is also relevant for consideration of changes in the standard.

Our goal was to test the robustness of the association between chronic exposure to $PM_{2.5}$ and mortality observed in the original study (Dockery et al. 1993), and the first extended follow-up of the Harvard Six Cities study (Laden et al. 2006) by replicating the analyses using 11 additional years of follow-up with exposures well below the U.S. annual standard (15 μg/m³) (U.S. EPA 1997). We examined different lags of exposure, tested the shape of the $PM_{2.5}$ concentration–mortality relationship, tested for changes in this slope over time, and relaxed the proportion assumption by allowing the effects of covariates to vary each year. We reexamined the association of $PM_{2.5}$ with specific causes of death such as

lung cancer and examined the effects of $PM_{2.5}$ depending on participants' chronic conditions and smoking status.

Methods

Study population. The Harvard Six Cities study population has been previously described (Dockery et al. 1993). Briefly, adults were randomly sampled from six cities in the eastern and midwestern United States between 1974 and 1977: in 1974, Watertown, Massachusetts; in 1975, Kingston and Harriman, Tennessee, and specific census tracts of St. Louis, Missouri; in 1976, Steubenville, Ohio, and Portage, Wyocena, and Pardeeville, Wisconsin; and in 1977, Topeka, Kansas. Information on age, sex, weight, height, educational level, smoking history, hypertension, and diabetes was collected by questionnaire at enrollment. All participants underwent spirometry tests at enrollment (Dockery et al. 1985) and chronic obstructive pulmonary disease (COPD) was defined as having

$$(FEV_1 \div FVC) < 70\%,$$

where FEV_1 is forced expiratory volume in 1 sec, and FVC is forced vital capacity. This analysis, as in the previous analyses, was restricted to 8,096 white participants with acceptable pulmonary function measurements. The study was approved by the Harvard School of Public Health Human Subjects Committee and all participants signed an informed consent before participation.

Mortality follow-up. Vital status and cause of death were determined by searching the National Death Index (NDI) for calendar years 1979–2009. Deaths before the NDI started in 1979 were identified by next of kin and Social Security records, and the cause of death was determined by a certified nosologist

who reviewed death certificates (Dockery et al. 1993).

Survival time. Survival times were calculated from enrollment until death or the end of follow-up (31 December 2009). For the 6 participants who were lost to follow-up before 1979, the censored survival times were calculated from enrollment to date of the last follow-up contact plus 6 months or the first day of the NDI (1 January 1979), whichever came first. For each cause of death category, participants who died from another cause were censored at time of death.

Air pollution estimates. Annual $PM_{2.5}$ concentration was assigned for each participant until death or censoring. $PM_{2.5}$ concentration was measured in the participant's city by a centrally located monitor from 1979 to 1986–1988, depending on the city (Dockery et al. 1993). Therefore, the study has no spatial contrast on the within-city scale. $PM_{2.5}$ concentrations for the years before monitoring started were assumed to be equal to the earliest monitored year. From the end of monitoring until 1998, $PM_{2.5}$ concentration was estimated from PM_{10} (aerodynamic diameter < 10 μm) data from U.S. EPA monitors and visibility (extinction) data from the National Weather Service (Laden et al. 2006). From 1999 through 2009, direct measurements of $PM_{2.5}$ were available from U.S. EPA monitors. For sensitivity analyses, we also predicted $PM_{2.5}$ for 1999–2009 (correlation between predicted and measured was 0.97) using the formula applied to derive exposure estimates during the earlier period when $PM_{2.5}$ was not measured.

Statistical analysis. We first replicated the original analysis separately for all-cause mortality, cardiovascular mortality as coded by the *International Classification of Diseases, 9th Revision* [ICD-9; World Health Organization (WHO) 1977] or the *10th Revision* (ICD-10; WHO 1992), 400.0–440.9, I10.0–I70.9, respectively, lung-cancer mortality (ICD-9 162, ICD-10 C33.0–C34.9), and COPD mortality (ICD-9 490.0–496.0, ICD-10 J40.0–J47.0) for the 36-year follow-up from 1974 to 2009 using a Cox proportional hazards model with follow-up time as the time scale (Dockery et al. 1993; Laden et al. 2006). $PM_{2.5}$ was included in each model as an annual time-dependent variable. The model was stratified by sex, age (1-year intervals) and time in the study (1-year intervals), so that each age/sex group had its own baseline hazard for each year of follow-up. The analysis was adjusted for potential confounders collected at baseline: smoking status (never, former, current), cumulative smoking (pack-years included separately for current and former smokers), educational level (< high-school, ≥ high school), and a linear and quadratic term for body mass index (BMI; kilograms per meter squared), using

the Cox proportional hazards model formulated as follows:

$$h_{is}(t) = h_{0,s}(t) \exp[\beta_1 X_i + \beta_2 Z_i(t)], \qquad [1]$$

where h_i is the instantaneous hazard probability of death for subject i in stratum s (defined by sex, age, and time in the study), $h_{0,s}(t)$ is the baseline hazard function, X_i is the vector of time-independent variables, and $Z_i(t)$ is the vector of time-dependent variables. We evaluated models with 1-year (i.e., exposure during the year before death or censure) to 5-year lagged moving averages and chose the best fit model using Akaike's information criterion (AIC) (Akaike 1973). The best fit moving average was determined from participants who survived at least 5 years from enrollment, so that AIC criteria were evaluated among populations with comparable sizes. We then estimated mortality rate ratios (RR) associated with $PM_{2.5}$ exposure during the best fit moving average on the whole sample size. Once the best exposure window was determined, we fit a penalized spline model using a cubic regression spline with 12 knots to estimate the shape of the concentration–response relation, and chose the optimal degree of freedom by minimizing AIC and evaluated nonlinearity with a Wald test. We investigated whether $PM_{2.5}$ advanced date of death for participants with chronic conditions at enrollment. We also investigated the potential for effect modification of $PM_{2.5}$ on mortality by smoking status at enrollment using interaction terms between such variables and $PM_{2.5}$. Finally, we tested the hypothesis that the effect of $PM_{2.5}$ changed over time by dividing the follow-up into four equally spaced time periods and testing interactions between period and $PM_{2.5}$.

Sensitivity analyses. We performed sensitivity analyses using a second-degree polynomial distributed lag model to allow the effects of $PM_{2.5}$ exposure to be distributed from 1 to 5 years before death or censor (Lepeule et al. 2006; Schwartz 2000); using predicted $PM_{2.5}$ concentrations after 1999 instead of the measured $PM_{2.5}$; considering only deaths from natural causes, with external causes of deaths (ICD-9 E800–E999, ICD-10 S00–T88 and V00–Y99) being censored at time of death; and considering only deaths that occurred in the state where the participants lived at enrollment. We next investigated the robustness of the results to alternative modeling assumptions by using a Poisson model with dummy variables for each year of follow-up, which is equivalent to a piecewise exponential proportionate hazard model with the baseline hazard changing each year (Laird and Oliver 1981):

$$\log \mu_{it} = \log E_{it} + \gamma_t T_t + \beta_1 X_i + \beta_2 Z_i(t), \quad [2]$$

where μ_{it} is the expected value of the death indicator for subject i at time t, E_{it} is the

exposure duration of subject i at time t (log E_{it} being the offset), T_t is the vector of dummy variables for time by 1 year (piecewise baseline hazard), X_i is the vector of the time-independent covariates, and $Z_i(t)$ is the vector of time-dependent variables. Using this Poisson survival analysis, we first compared the results to the Cox model and then relaxed the proportionate hazard assumption for sex, education, and cumulative smoking by including interaction terms of these variables with each year of follow-up. As an alternative to the previous analyses (Dockery et al. 1993; Laden et al. 2006), we used age in 5-year groups as the time scale, and adjusted the model for time trends (linear term). For specific causes of death, convergence issues led us to group age by 10 years. We then fit penalized spline models. Because RRs may vary over time and period-specific RRs may be biased, we used the Poisson model to calculate adjusted survival curves (Hernan 2010). We included product terms between $PM_{2.5}$ and time in model 2 [Equation 2], thereby allowing the effect of $PM_{2.5}$ to flexibly vary from year to year. We then predicted the survival probability for each year of follow-up for each participant under three scenarios using concentrations of $PM_{2.5}$ throughout the entire follow-up period equal to 10, 15, or 20 μg/m³.

p-Values < 0.05 were considered statistically significant. All analyses were repeated separately for all- and specific-causes of deaths. Analyses were conducted with SAS software, version 9.2 (SAS Institute Inc., Cary, NC) and R statistical software, version 2.12.2 (R Foundation for Statistical Computing, Vienna, Austria).

Results

Study population. The 8,096 participants were 25–74 years of age at enrollment (mean ± SD, 49.6 ± 13.4) and 54.7% were female. More than half of the participants had a high school degree or higher, 35.8% were current smokers, and 23.9% were former smokers and the average BMI was 25.8 ± 4.5. As for chronic conditions, 17.8% reported hypertension, 11.6% COPD, and 6.9% diabetes.

Mortality rates and $PM_{2.5}$ levels. At the end of 2009, there were 212,067 person-years of follow-up and 55.5% of the participants had died, among whom 40.8% died from cardiovascular diseases, 7.8% from lung cancer, and 5.5% from COPD (Table 1). Overall, $PM_{2.5}$ concentration decreased during the study period (Figure 1). After 1998, annual average levels declined by 1.8 μg/m³ in St. Louis and by 10.5 μg/m³ in Steubenville, whereas levels increased by 1.5 μg/m³ in the Portage–Wyocena–Pardeeville area. Since 2000, all the cities experienced average $PM_{2.5}$ levels < 15 μg/m³ except Kingston–Harriman and Steubenville, which had average concentrations of ≤ 18 μg/m³.

Association between PM$_{2.5}$ and mortality. Using the Cox proportional hazards model, statistically significant associations between PM$_{2.5}$ exposure and all-cause, cardiovascular, and lung-cancer mortality were observed (Table 2). The AIC indicated lag 1 (i.e., exposure during the previous year) to be the best fit exposure window for all-cause mortality [see Supplemental Material, Table 1 (http://dx.doi.org/10.1289/ehp.1104660)]. For cause-specific mortality, the best fit moving average differed between the Cox and the Poisson regressions. Because the differences in AIC were very small between the 1- and 5-year moving averages for both the Cox and Poisson regressions, we chose the longer of the two moving averages to produce more stable results, specifically, a 1- to 3-year moving average for cardiovascular and lung-cancer mortality, and a 1- to 5-year moving average for COPD mortality. Each 10-µg/m^3 increase in PM$_{2.5}$ was associated with a 14% increased risk of all-cause death [95% confidence interval (CI): 7%, 22%], a 26% increase in cardiovascular death (95% CI: 14%, 40%), and a 37% increase in lung-cancer death (95% CI: 7%, 75%). For both all-cause mortality and

specific causes of death, the model fit was better without the spline (*p*-values between 0.24 and 0.43), indicating a linear relationship with PM$_{2.5}$. Results restricted to participants with chronic conditions at enrollment (i.e., hypertension, COPD, or diabetes) were consistent with those estimated for all participants (Table 2). Although, the interaction between smoking status and PM$_{2.5}$ was not statistically significant, there was a trend for a stronger estimated effect of PM$_{2.5}$ on mortality in current and former smokers. However, positive associations between PM$_{2.5}$ and all-cause and cardiovascular mortality were still evident in never smokers. RR for PM$_{2.5}$ fluctuated over time for all-cause mortality and specific causes of death, without clear trends (Table 2).

Sensitivity analysis. For both all causes and specific causes of death, the cumulative effects estimated from the polynomial distributed lag model were similar to the effect estimates obtained with the selected moving averages (Table 2). However, the five lags were too correlated (between 0.90 and 0.96) to disentangle the relative importance of each one. Using predicted PM$_{2.5}$ instead of measured PM$_{2.5}$ for exposures after 1999, excluding the

138 deaths from external causes and excluding the 702 participants who died in a state other than the state where they lived at enrollment, did not change the results (data not shown) except for the lung-cancer mortality association with PM$_{2.5}$, which was slightly attenuated (increased risk of 28%; 95% CI: –2%, 67% compared with 37%; 95% CI: 7%, 75%) when the 702 participants were excluded.

With the Poisson framework, using basic assumptions, relaxed proportionate hazard assumption for covariates, or age as the time scale, the effect estimates and *p*-values fluctuated slightly but without any notable change in the results compared with estimates from the Cox models for all-cause mortality and for specific-causes of death (data not shown). The penalized spline models also indicated linear concentration–response relationships without a threshold for PM$_{2.5}$ and mortality from all-causes and specific-causes [see Supplemental Material, Figure 1 (http://dx.doi.org/10.1289/ehp.1104660)]. With the Poisson survival analysis, we predicted survival assuming every participant was exposed to a constant concentration of PM$_{2.5}$ (10, 15, or 20 µg/m^3) during the entire follow-up period. Adjusted

Table 1. Number of participants, mortality, and average PM$_{2.5}$ levels in the Harvard Six Cities study, 1974–2009.

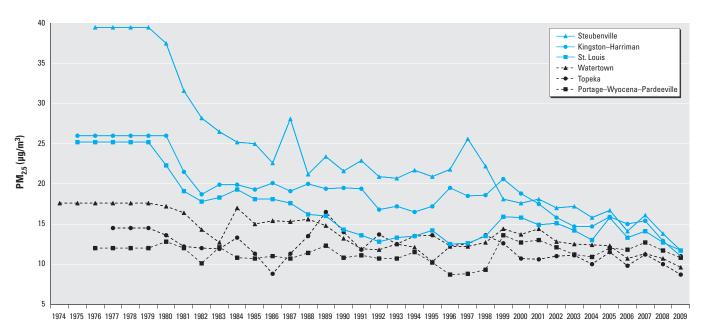| Characteristic	Six cities (combined)	Steubenville	Kingston–Harriman	St. Louis	Watertown	Topeka	Portage–Wyocena–Pardeeville
Participants (*n*)	8,096	1,346	1,258	1,292	1,332	1,238	1,630
Person-years (*n*)	212,067	33,276	33,067	32,225	36,818	32,877	43,804
Cause of death							
All causes [*n* (%)]	4,495 (55.5)	822 (61.1)	733 (58.3)	827 (64.0)	700 (52.6)	617 (49.8)	796 (48.8)
Cardiovascular (%)	40.8	45.3	41.1	42.2	39.3	37.4	38.6
Lung cancer (%)	7.8	9.0	8.0	8.7	6.6	7.3	6.8
COPD (%)	5.5	4.9	7.0	5.1	4.9	7.3	4.6
1974–2009 average of individual PM$_{2.5}$ concentrations	15.9	23.6	19.1	16.7	14.0	12.2	11.4

Figure 1. Annual mean PM$_{2.5}$ levels during 1974–2009 in the Harvard Six Cities study.

for individual covariates, the lowest $PM_{2.5}$ concentration was associated with the highest survival (Figure 2). The three adjusted survival curves showed that the proportionate hazard was a reasonable assumption for $PM_{2.5}$ and that $PM_{2.5}$ effects were quite stable over time.

Discussion

Including more recent observations with $PM_{2.5}$ exposures down to 8 μg/m³, we continued to find a statistically significant association between chronic exposure to $PM_{2.5}$ and all-cause and cardiovascular mortality. Furthermore, in the present extended follow-up, $PM_{2.5}$ exposure was also statistically significantly associated with lung-cancer mortality. Our study indicated no sensitivity of

the results for all-cause mortality and specific causes of death when we allowed the effects of smoking, education, and sex to vary over time, or when we used age as the time scale instead of follow-up time. Using very flexible modeling assumptions, our results did not show any rationale for change of $PM_{2.5}$ effect size over the whole study period, as indicated by the adjusted survival curves and the lack of a clear interaction of $PM_{2.5}$ with the four study periods. The concentration–response relationship was linear without any threshold, even at exposure levels below the U.S. annual 15-μg/m³ standard (U.S. EPA 1997). Taken together with the results of a previous reanalysis of the Harvard Six Cities study (Krewski et al. 2005b), there is evidence for a robust

association between chronic $PM_{2.5}$ exposure and early mortality.

Consistency of the results. Our results indicated a statistically significant 14% increase in all-cause mortality for a 10-μg/m³ annual increase in $PM_{2.5}$, which is similar to the results of the previous follow-ups (Dockery et al. 1993; Laden et al. 2006). The Netherlands Cohort Study on Diet (NLCS–Air) in Europe (Beelen et al. 2008b), the Adventist Study (McDonnell et al. 2000), and the male Health Professionals Follow-up Study in the United States (Puett et al. 2011) did not show statistically significant associations between $PM_{2.5}$ and all-cause mortality. However, our current results are consistent with those from the ACS cohort (Pope et al. 2002), the Nurses' Health Study (Puett et al. 2009), and the Medicare cohort (Eftim et al. 2008), which indicated mortality increases ranging from 3–26% per 10-μg/m³ increase in $PM_{2.5}$.

The 26% increase in cardiovascular mortality for each 10-μg/m³ increase in $PM_{2.5}$ exposure during the previous 3 years estimated in this extended follow-up is similar to the previous estimate (Laden et al. 2006). Although the NLCS–Air study (Beelen et al. 2008b) found no statistically significant association, the magnitude of the estimated effect reported here is between the 12% increase estimated for the ACS cohort (Pope et al. 2004) and the 76% increase estimated for the Women's Health Initiative study (Miller et al. 2007). Puett et al. (2009) also estimated a 100% increase in fatal coronary heart diseases for a 10-μg/m³ increase in $PM_{2.5}$ during the prior year. Underlying mechanisms for the effects of $PM_{2.5}$ on cardiovascular mortality are still poorly understood, but changes in vasoconstriction might explain the associations (Anderson et al. 2011).

The previous extended follow-up of the Harvard Six Cities study showed an elevated, but not statistically significant, risk of lung-cancer mortality (Laden et al. 2006), whereas the present extended follow-up estimated a statistically significant 37% increase in lung-cancer mortality (for each 10-μg/m³ increase in $PM_{2.5}$), which is greater than that estimated for both the ACS cohort (14%) (Pope et al. 2002) and a Japanese cohort (27%) (Katanoda et al. 2011). Lungs are one of the organs that are most directly affected by particulate air pollution. Fine particles, which may carry toxic chemicals of carcinogenic potential (Laden et al. 2000), can reach lung alveoli where the clearance is slow (Pinkerton et al. 1995) and induce durable pulmonary and systemic inflammation (Riva et al. 2011). Recent findings in the ACS cohort indicated that a 10-μg/m³ increase in $PM_{2.5}$ concentration was associated with a statistically significant 15% to 27% increase in lung-cancer mortality in never smokers (Turner et al. 2011).

Table 2. Adjusted[a] association between $PM_{2.5}$[b] and mortality, for the 8,096 participants and certain subpopulations of the Harvard Six Cities study, 1974–2009.

Cause of death/stratum-specific estimates according to characteristics at enrollment	n participants (n person-years)	RR (95% CI) for 10-μg/m³ increase in $PM_{2.5}$
All-cause	8,096 (212,067)	1.14 (1.07, 1.22)
Chronic conditions[c]		
Hypertension	1,439 (30,540)	1.17 (1.03, 1.32)
COPD	942 (17,723)	1.09 (0.95, 1.26)
Diabetes	563 (11,473)	1.04 (0.85, 1.27)
Smoking status (p-interaction = 0.58)		
Never smoker	3,265 (90,372)	1.09 (0.98, 1.21)
Former smoker	1,934 (48,049)	1.17 (1.04, 1.30)
Current smoker	2,897 (73,646)	1.17 (1.06, 1.28)
Follow-up period (p-interaction = 0.06)		
1974–1982	8,096 (58,798)	1.06 (0.96, 1.17)
1983–1991	7,478 (63,129)	1.32 (1.16, 1.50)
1992–2000	6,391 (51,800)	1.11 (0.98, 1.27)
2001–2009	4,910 (38,340)	1.19 (0.91, 1.55)
Cardiovascular	7,961 (195,941)	1.26 (1.14, 1.40)
Smoking status (p-interaction = 0.45)		
Never smoker	3,232 (83,861)	1.21 (1.04, 1.41)
Former smoker	1,891 (44,205)	1.21 (1.02, 1.44)
Current smoker	2,838 (67,875)	1.36 (1.17, 1.58)
Follow-up period (p-interaction = 0.07)		
1974–1982	7,961 (42,672)	1.08 (0.92, 1.27)
1983–1991	7,478 (63,129)	1.46 (1.21, 1.76)
1992–2000	6,391 (51,800)	1.30 (1.06, 1.59)
2001–2009	4,910 (38,340)	1.57 (1.01, 2.43)
Lung cancer	7,961 (195,941)	1.37 (1.07, 1.75)
Smoking status (p-interaction = 0.15)		
Never smoker	3,232 (83,861)	1.25 (0.54, 2.89)
Former smoker	1,891 (44,205)	1.96 (1.29, 2.99)
Current smoker	2,838 (67,875)	1.25 (0.95, 1.64)
Follow-up period (p-interaction = 0.19)		
1974–1982	7,961 (42,672)	1.45 (0.98, 2.15)
1983–1991	7,478 (63,129)	0.94 (0.58, 1.52)
1992–2000	6,391 (51,800)	1.54 (0.98, 2.41)
2001–2009	4,910 (38,340)	2.84 (1.06, 7.59)
COPD	7,805 (180,106)	1.17 (0.85, 1.62)
Smoking status (p-interaction = 0.35)		
Never smoker	3,191 (77,422)	0.85 (0.36, 2.02)
Former smoker	1,847 (40,453)	1.64 (0.92, 2.93)
Current smoker	2,767 (62,231)	1.10 (0.74, 1.62)
Follow-up period (p-interaction = 0.35)		
1974–1982	7,805 (26,837)	0.79 (0.36, 1.72)
1983–1991	7,478 (63,129)	1.52 (0.90, 2.56)
1992–2000	6,391 (51,800)	1.31 (0.74, 2.31)
2001–2009	4,910 (38,340)	0.68 (0.25, 1.83)

[a]Cox proportional hazards model stratified by sex, age, and time in the study and adjusted for BMI, education, and smoking history. [b]$PM_{2.5}$ moving average was 1 year before death or censure for all-cause deaths, 1–3 years for cardiovascular and lung-cancer deaths, and 1–5 years for COPD deaths. [c]Estimates restricted to participants with the specified chronic condition.

We did not find such an association in our study, which might be due to a lack of statistical power (350 lung-cancer deaths, 26 among never smokers). However, estimated effects of PM$_{2.5}$ on all-cause and cardiovascular mortality were also statistically significant (or borderline significant) in never smokers, and higher in current smokers compared to never or former smokers (Table 2).

Regarding COPD mortality, we found a positive but not statistically significant risk of COPD death associated with PM$_{2.5}$ exposure. In the ACS cohort, Pope et al. (2004) estimated an unexpected inverse association between PM$_{2.5}$ exposure and COPD mortality, whereas Katanoda et al. (2011) estimated an inverse but not statistically significant association between PM$_{2.5}$ and COPD in a Japanese cohort.

Chronic conditions at enrollment and mortality. The central deposition of particles in lungs has been shown to be enhanced in COPD patients (Bennett et al. 1997). Although PM$_{2.5}$ has been associated with early mortality in COPD patients (Zanobetti et al. 2008), and ozone has been associated with early mortality in susceptible subjects (i.e., with COPD, diabetes, heart failure, or myocardial infarction) (Zanobetti and Schwartz 2011), our results did not indicate stronger associations in participants with such chronic conditions at enrollment compared with the population as a whole,. This might have been due to a lack of statistical power as few participants had COPD (*n* = 942) or diabetes (*n* = 563) at enrollment.

Exposure assessment. Use of outdoor measurements from central monitoring stations as a proxy measure of mean personal exposure to PM$_{2.5}$ is prone to measurement error because the measures do not capture fine spatial contrasts that may occur within a city, which may bias the results. Recent reanalyses of the ACS cohort using land use regression models showed that the impact on the PM$_{2.5}$–mortality association was heterogeneous depending on the city (Krewski et al. 2009). However, other recent studies have suggested that considering a more precise exposure model focused on the home address might not improve health effects estimates in terms of bias and variance (Kim et al. 2009; Lepeule et al. 2010; Szpiro et al. 2011). In the Harvard Six Cities study, there were not enough monitors in the cities to implement a land use regression model.

Strengths and limitations. Our results were adjusted for baseline factors, but there is potential for residual confounding for risk factors after enrollment and for unmeasured factors such as occupational exposures or medication use if those factors co-vary with PM$_{2.5}$. Some other limitations are that we did not measure PM$_{2.5}$ in the same locations throughout the study period, that death certificates might have listed misclassified specific causes of death, and that hypertension and diabetes were assessed by questionnaire only. An extensive body of methodological work has been performed regarding the sensitivity of estimated associations between long-term exposure to air pollution and mortality, especially for the ACS and Harvard Six Cities study cohorts. More specifically, it has been shown that results were robust to alternative model specifications, alternative metrics of PM$_{2.5}$, and adjustment for individual and ecological risk factors such as occupational exposures and socioeconomic variables (Krewski et al. 2005a, 2005b). It was also shown that using a spatial covariance structure did not change the results (Pope et al. 2002), but with only six locations, that methodology is not applicable in our study. Whereas the primary analysis from the Harvard Six Cities study (Dockery et al. 1993) estimated associations were based on between–city contrasts in exposure, in the current study, with age used as time scale, the exposure relied on both between– and within–city contrasts, limiting the potential for residual cross-sectional confounding. The strengths of the present study are the randomly sampled participants and its extended follow-up through 2009, which included more observations of participants with lower exposures during recent years and provided more statistical power.

Critical periods of PM$_{2.5}$ exposure. Our results indicated that the best fit moving average for PM$_{2.5}$ was 1 year for all-cause mortality. For cardiovascular and lung-cancer mortality, no clear pattern was identified because of the high correlation between PM$_{2.5}$ concentrations in the 5 lagged years tested. These results suggest that PM$_{2.5}$ exposure can act to promote cardiovascular diseases and lung-cancer growth, although the design of this study precludes us from determining whether PM$_{2.5}$ initiates these diseases as suggested by other studies (Beelen et al. 2008a; Beeson et al. 1998). These results agree with the literature (Gehring et al. 2006; Krewski et al. 2009; Puett et al. 2009; Schwartz et al. 2008) and suggest that health improvements can be expected almost immediately after a reduction in air pollution. This conclusion should be taken into account for cost–benefit analyses related to air pollution standards.

Role of sulfates and public health implications. Although RRs for PM$_{2.5}$ fluctuated over time, our extended follow-up did not indicate any clear pattern over time during the study period. Between 1979–1988 (Laden et al. 2000) and 2009 (Nehls and Akland 1973), the sulfates/PM$_{2.5}$ ratio for exposures measured for the Harvard Six Cities study dropped between 13% and 54%, depending on the city. If sulfates are unrelated to mortality, as some have argued (Grahame and Schlesinger 2005), the elimination of a substantial fraction of nontoxic material from PM$_{2.5}$ mass should result in a substantial increase in the PM$_{2.5}$ coefficient, which would otherwise have been suppressed by the large fraction of mass that was nontoxic. This was not the case, and hence our results indicate that sulfate particles are about as toxic as the average fine particle. This is consistent with the results of Pope et al. (2007), who found that the 2.5-μg/m^3 decrease in sulfate particle concentrations observed during an 8-month smelters strike were associated with a 2.5% decrease in the number of deaths in the region. In comparison, a 2.5-μg/m^3 decrease in PM$_{2.5}$ in our follow-up of the Harvard Six Cities study was associated with a 3.5% reduction in all-cause deaths, but that was for reductions in PM$_{2.5}$ lasting at least a year, not
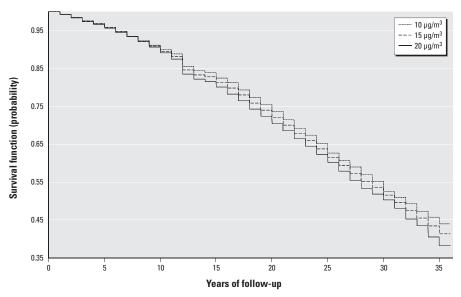

Figure 2. Survival probabilities under three hypothetical scenarios: participants of the Harvard Six Cities study are exposed to 10, 15, or 20 μg/m^3 PM$_{2.5}$ during the entire follow-up period of 1974–2009.

8 months. Given that there were 2,423,712 deaths in the United States in 2007 (Xu et al. 2010) and that the average PM$_{2.5}$ level was 11.9 µg/m^3 (U.S. EPA 2011), our estimated association between PM$_{2.5}$ and all-cause mortality implies that a decrease of 1 µg/m^3 in population-average PM$_{2.5}$ would result in approximately 34,000 fewer deaths per year.

Conclusion

Including recent observations with PM$_{2.5}$ exposures well below the U.S. annual standard of 15 µg/m^3 and down to 8 µg/m^3, the relationship between chronic exposure to PM$_{2.5}$ and all-cause, cardiovascular, and lung-cancer mortality was found to be linear without a threshold. Our results were not sensitive to various model specifications. Furthermore, estimated effects of PM$_{2.5}$ did not change over time, suggesting a stable toxicity of PM$_{2.5}$, even at lower exposure levels and with a lower sulfates proportion. These results suggest that further public policy efforts that reduce fine particulate matter air pollution are likely to have continuing public health benefits.

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