Estimating the Causal Effect of Fine Particulate Matter Levels on Death and Hospitalization: Are Levels Below the Standards Harmful?

Supplementary material

Summary statistics

Table S1: Mean and standard deviation of all variables (potential confounders, p=122) and of the four health outcomes available in the study. Calculations shown are for the full cohort and computed over all observations (total = $68,789$). Study period 2002 to 2010. Variables from US census and USDA are available at the area level (zip codes and county) and mean and standard deviations are calculated across these area-level summaries.

Propensity score overlap

To ensure that the distribution of the covariates are overlapping in the region of high ad low pollution, we have plotted a histogram of the estimated propensity scores in both the high (> 12) and low pollution (< 12) groups, respectively (see Figure **S1**). We found that there is adequate overlap between these two distributions suggesting that positivity assumption is not violated in our study.

Figure S1: Distribution of the estimated propensity scores for the high and low exposure groups

Sensitivity Analyses

Main Analysis: Inverse probability weighting with Binary Exposure: In the main analyses we presented results by using inverse probability weighting (IPW). IPW estimates the effect of the exposure by re-weighting our sample using the stabilized inverse probability weights (Table 2, and Figure 4).

Sensitivity Analyses

SA1: Continuous exposure model

We also fitted to the data a Cox proportional hazard model with a linear E-R function, where the $p=122$ potential confounders are included into the model as linear terms. We define the continuous exposure as average of PM2.5 two years prior the interview date.

Table S3 shows the estimated hazard ratios and confidence intervals. Hazards ratio estimates are calculated as exponents of the coefficients and confidence intervals are calculated based on robust variances accounting for multiple observations per patient. Our estimate for the effect of $PM_{2.5}$ on all cause mortality is comparable to the same estimate in Table 2 of the ACS although it is not statistically significant. The ACS study reports an estimated HR of 1.06 (1.02, 1.11) associated with a 10 μ g/m³ increase in PM_{2.5.} This is largely consistent with our findings in light of differences between the two studies as outlined in Table 1 in the main text. For example, since the ACS study was conducted at an earlier time period with higher average PM2.5 and a larger study population, it does not come as a surprise that their estimate is significant while ours is not.

Table S3: Hazard ratios showing the effect of an increase of 10 μ g/m³ increase in PM_{2.5.} obtained from fitting a CPH with the average exposure to $PM_{2.5}$ as the main exposure variable. Table reports 95% confidence intervals based on robust, sandwich variance estimators (Results of SA1).

Figure S2 shows the CPH estimates using the linear exposure term and after direct adjustment for confounders including and excluding MCBS variables and is similar to Figure 2 in the main analysis. One again we find that the results are consistent with the main analysis, suggesting that our estimates are robust to the exclusion of the MCBS variables into the model to adjust for confounding.

Figure S2: shows the estimated hazard ratios and 95% CI obtained by fitting a Cox Proportional Hazard Model (CPH) with a continuous exposure and confounding adjustment obtained by including the covariates linearly into the model. Estimates in blue are obtained by excluding from the model the MCBS variables. (Results of SA1)

SA2: Continuous Exposure and Cox Proportional Hazard Model with Non Linear Exposure-Response (ER) function

Table S4 shows the p-value of the Wald test for nested models testing as whether globally there is evidence of non-linearity in the ER function. P-values suggest that there might be a nonlinear relationship between exposure to PM2.5 and circulatory hospitalizations in the low pollution cohort. Further investigations of this relationship are left for future work (Results for SA2).

Table S4: Wald test for the significance of the spline parameters

Figure S3: Shows the exposure-response curve for PM2.5 and the four outcomes looked at. The nonlinear curve was fit using splines with 3 degrees of freedom. (Results of SA2)

SA3: Low pollution cohort using cutoff of 10 μ g/m³. This analysis again restricts subjects to those living in areas lower than $12 \mu g/m^3$, but now defines the binary exposure to be an indicator whether a subject lives in an area with average pollution levels below 10 μ *g/m³* instead than 8 μ *g/m³* as in the main analysis.

Table S5: Effect estimates for low pollution cohort using the exposure to be an indicator that PM_{2.5} is below 10 μ *g/m³*.

	Low pollution cohort using WHO cutoff, $N = 18,144$ person years = $34,429$
All cause mortality	1.09(0.98, 1.22)
All cause hospitalization	1.03(0.98, 1.08)
Circulatory hospitalization	1.03(0.98, 1.09)
Respiratory hospitalization	1.04(0.97, 1.12)

Figure S4: shows the estimated hazard ratios and 95% CI obtained by fitting a Cox Proportional Hazard Model (CPH) with a binary exposure and confounding adjustment done via inverse probability weighting. Estimates in blue are obtained by excluding from the PS model the MCBS variables. (Results of SA3)

