Long-Term Exposure to Air Pollution and Incidence of Myocardial Infarction: A Danish Nurse Cohort Study

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BACKGROUND: Air pollution exposure has been linked to coronary heart disease, although evidence on PM2.5 and myocardial infarction (MI) incidence is mixed.

OBJECTIVES: This prospective cohort study aimed to investigate associations between long-term exposure to air pollution and MI incidence, adjusting for road traffic noise.

METHODS: We used data from the nationwide Danish Nurse Cohort on 22,882 female nurses (>44 years of age) who, at recruitment in 1993 or 1999, reported information on cardiovascular disease risk factors. Data on MI incidence was collected from the Danish National Patient Register until the end of 2014. Annual mean concentrations of particulate matter (PM) with a diameter <2.5 μg/m3 (PM2.5), PM10, nitrogen dioxide (NO2), and nitrogen oxides (NOx) at the nurses’ residences since 1990 (PM10 and PM2.5) or 1970 (NO2 and NOx) were estimated using the Danish Eulerian Hemispheric Model/Urban Background Model/AirGIS (DEHM/UBM/AirGIS) dispersion model. We used time-varying Cox regression models to examine the association between 1- and 3-y running means of these pollutants, as well as 23-y running means of NO2 and NOx, with overall and fatal incident MI. Associations were explored in three progressively adjusted models: Model 1, adjusted for age and baseline year; Model 2, with further adjustment for potential confounding by lifestyle and cardiovascular disease risk factors; and Model 3, with further adjustment for road traffic noise, modeled as the annual mean of a weighted 24-h average (Lden).

RESULTS: Of the 22,882 women, 641 developed MI during a mean follow-up of 18.6 years, 121 (18.9%) of which were fatal. Reported hazard ratios (HRs) were based on interquartile range increases of 5.3, 5.5, 8.1, and 11.5 μg/m3 for PM2.5, PM10, NO2, and NOx, respectively. In Model 1, we observed a positive association between a 3-y running mean of PM2.5 and an overall incident MI with an HR = 1.20 (95% CI: 1.07, 1.35), which attenuated to HR = 1.06 (95% CI: 0.92, 1.23) in Model 2. In Model 1 for incident fatal MI, we observed a strong association with a 3-y running mean of PM2.5, with an HR = 1.69 (95% CI: 1.33, 2.13), which attenuated to HR = 1.35 (95% CI: 1.01, 1.81) in Model 2. Similar associations were observed for NO2 and NOx. No evidence of an association was observed for NO2 or NOx. For all pollutants, associations in Model 2 were robust to further adjustment for road traffic noise.

CONCLUSIONS: We found no association between long-term exposure to PM2.5, PM10, NO2, or NOx and overall MI incidence, but we observed positive associations for PM2.5 and PM10 with fatal MI. We present novel findings that the association between PM and MI incidence is robust to adjustment for road traffic noise.

Introduction

There is growing evidence on the association between long-term exposure to air pollution and cardiovascular disease (CVD) mortality (Faustini et al. 2014; Lelieveld et al. 2019; Pope et al. 2004). Investigations into the association between air pollution and the incidence of specific cardiovascular events, such as myocardial infarction (MI), however, are inconsistent and still under debate. A number of studies have detected significant positive associations between MI incidence and air pollutants such as nitrogen dioxide (NO2) (Grazuleviciene et al. 2004; Rosenlund et al. 2009; Roswall et al. 2017) and particulate matter (PM) with a diameter <2.5 μg/m3 (PM2.5) (Bai et al. 2019a; Danesh Yazdi et al. 2019; Clompmaker et al. 2019; Madrigano et al. 2013; Yusuf et al. 2020), <10 μg/m3 (PM10) (Rosenlund et al. 2009), or <0.1 μg/m3 [ultrafine particles (UFPs)] (Bai et al. 2019b; Downward et al. 2018). However, in many of the aforementioned, and other, studies, associations for NO2 (Atkinson et al. 2013; Bai et al. 2019a, 2019b; Downward et al. 2018; Lipsett et al. 2011; Rosenlund et al. 2006), PM2.5 (Downward et al. 2018; Lipsett et al. 2011; Miller et al. 2007; Puett et al. 2011, 2009), PM10 (Atkinson et al. 2013; Downward et al. 2018; Klompmaker et al. 2019; Lipsett et al. 2011; Puett et al. 2011, 2008; Rosenlund et al. 2006), and nitrogen oxides (NOx) (Bodin et al. 2016; Downward et al. 2018; Lipsett et al. 2011; Stockfelt et al. 2015) have been null or nonsignificant. Notably, those studies that examined nonfatal and fatal MI incidence separately reported stronger associations with fatal MI for NO2 (Rosenlund et al. 2009, 2006; Roswall et al. 2017) and PM10 (Rosenlund et al. 2009, 2006). This raises questions about the
mechanisms underlying air pollution–induced MI and may imply that air pollution leads to more severe, or fatal, ischemic events.

Recently, associations between road traffic noise exposure and both MI (Babisch 2014; Selander et al. 2013) and cardiovascular mortality (Heinonen-Guzejev et al. 2007) have been documented. Because road traffic is an important source of both urban air pollution and noise exposure, it is important to take potential confounding effects or effect modification by road traffic noise exposure into account when assessing the health effects of urban air pollution. Only three studies on air pollution and MI incidence have examined the potential confounding effects by road traffic noise (Bai et al. 2019b; Bodin et al. 2016; Roswall et al. 2017). Roswall et al. (2017) showed that NO2 and road traffic noise were both significantly associated with MI and that mutual exposure adjustment attenuated the association with NO2, but not with noise. Bai et al. (2019b) reported a weak but significant association between UFP exposure and MI that was unchanged following adjustment for road traffic noise, and Bodin et al. (2016) found no evidence of an association between noise or NOx with MI. Furthermore, in another recent nationwide Swiss study assessing associations with MI mortality, Héritier et al. (2019) reported that the effects of air pollution were attenuated after adjustment for transportation noise and concluded that air pollution studies not adequately accounting for the effects of transportation noise exposure risk overestimating the cardiovascular disease burden of air pollution.

To date, no single cohort study has assessed the effects of PM in single-pollutant models adjusted for the effects of road traffic noise. In the present study, we aimed to investigate the relationship between long-term exposure to PM2.5, PM10, NO2, and NOx and incidence of MI, separately for overall and fatal incident MI, while considering the potential attenuating effects of road traffic noise as a confounder.

Methods

The Danish Nurse Cohort

Established in 1993, and inspired by the American Nurses’ Health Study, the Danish Nurse Cohort was designed to examine the effects of hormone replacement therapy in the Danish population (Hundrup et al. 2012). The Danish Nurse Organization (DNO), comprising both actively working and retired Danish nurses, had a 95% inclusion rate at the start of the study period and is highly representative of the Danish nursing workforce. In 1993, initial questionnaires were mailed to all 23,170 female DNO members (>44 years of age), to which 19,898 (86%) responded. In 1999, the cohort was re-investigated and an additional 8,833 (69%) nurses (either newly invited women who turned 45 years of age in the interim, or re-invited non-responders from 1993) were included. Upon inclusion, information was collected on working conditions, lifestyle factors (including dietary, physical activity, smoking, and alcohol consumption habits), and various measures of self-reported health, including height, weight, history of various diseases (including cardiovascular), and reproductive health history. In 1999, additional information on lifetime changes in weight and family body type was gathered. All 28,731 female nurses received unique identification numbers, which were then linked to the Danish Civil Registration System (Pedersen 2011) for the collection of vital statistics (date of death or emigration, and full residential address history) between 1970 and whichever of the following came first: date of death, emigration, or disappearance or 31 December 2014.

Outcome Definitions

Information on MI incidence was obtained by linking the nurses to the Danish National Patient Registry (DNPR), and the Danish Register of Causes of Death (DRCD) between the baseline (year of cohort entry: 1 April 1993 or 1 April 1999) and end of follow-up on 31 July 2015. The DNPR collects data on both primary and secondary diagnoses affiliated with each hospital contact, registered upon either hospital discharge or completion of an outpatient visit. Established in 1977, and reaching nationwide coverage by 1978, the DNPR is updated continuously. Between 1982 and 2012, numerous studies have performed a medical record validation of MI as a primary diagnosis in the DNPR and have consistently reported positive predictive values (PPVs) above 90% (Joensen et al. 2009; Madsen et al. 2003; Sundbøll et al. 2016; Thygesen et al. 2011). In one study, two separate PPVs were calculated for best- and worst-case scenarios. This study reported a 100% PPV in the best-case scenario based only on assessable or retrievable cases and a 74% PPV in the worst-case scenario, for which their calculation included those cases deemed insufficient for validation (Coloma et al. 2013). In this study, we categorized MI according to the International Classification of Disease, Eighth Revision (ICD-8; WHO 1966) or the International Statistical Classification of Diseases and Related Health Problems, Tenth Revision (ICD-10; WHO 2016). MI (ICD-8, code 410; ICD-10 code I21) was defined as incident either if it was recorded as a first-time diagnosis between baseline and end of follow-up in the DNPR, or, if it was fatal, as an out-of-hospital MI death in the DRCD (prior to 31 December 2013) if there were no prior DNPR MI diagnoses. Fatal cases of incident MI further included all those who died from any cause within 28 d following the hospital admission date for an incident MI between baseline and the end of follow-up, as determined from linkage to the Civil Registration System.

Exposure Assignment

By linking the participants’ unique identification numbers, which are administered to all Danish residents through the Danish Civil Registration System, to the Danish Address Database, we collected the geographical coordinates of the participants’ residential addresses from 1971 through 31 December 2014. These coordinates were subsequently used to estimate the nurses’ residential exposures to both air pollution and road traffic noise.

Assessment of Residential Air Pollution Concentrations

Annual average exposure levels to PM2.5, PM10, NO2, and NOx were estimated using the Danish air pollution modeling system Danish Eulerian Hemispheric Model/Urban Background Model/AirGIS (DEHM/UBM/AirGIS): http://envis.au.dk/en/knowledge/airmodels/airgis/ (Jensen et al. 2017, 2001; Khan et al. 2019), comprising three air pollution models (Brandt et al. 2001). These three models include the DEHM (Brandt et al. 2012; Frohn et al. 2002), which accounts for the long-range transport components, the Danish UBM (Brandt et al. 2012; Frohn et al. 2002), which calculates the local background on a 1-km² resolution grid overlaying Denmark, and the Operational Street Pollution Model (OSPM) (Berkovicz 2000; Ketzel et al. 2012), which estimates the concentration at the residential address’ front door. DEHM/UBM/AirGIS is a validated air pollution exposure model, and the OSPM utilized for the street-pollution estimates has been employed in over 25 countries (Hvidtfeldt et al. 2018; Ketzel et al. 2012). Comparisons of model results with measurements from the Danish routine monitoring program (comprising 17 stations across the country) show $R^2$ values in the range of 0.7–0.9 depending on site, pollutant, and averaging period (Ellermann et al. 2020). The air pollution exposure modeling system undergoes smaller updates continuously, and for each update careful comparisons with all available measurement data are performed. This study used the annual mean concentrations of PM2.5, PM10
models: Model 1, adjusted for age (underlying timescale) and year of cohort entry (1993 or 1999); Model 2, further adjusted for potential confounders and CVD risk factors; and Model 3, further adjusted for continuous (per 10 dB) residential road traffic noise levels (i.e., L_{den}). Potential confounders and risk factors included smoking status (never/previous/current), number of pack-years, fruit consumption (rarely/few times per week/daily or several times per day), avoidance of fatty meat (yes/no), alcohol consumption in grams (g) per week (never: 0/moderate: 1–14/heavy: >15), physical activity (low/medium/high), marital status (married/separated/divorced/single/widowed), employment status (working/home-maker/retired/unemployed or in rehabilitation/other), use of oral contraceptives (ever/never), hormone replacement therapy use (never/previous/current), and level of urbanization (rural/provincial/urban). Level of urbanization was defined using the following population density thresholds: rural: <180 persons/km²; provincial: 180–185, 220 persons/km²; urban: >5,220 persons/km².

Effect modification of the association between the 3-y running mean of residential PM\textsubscript{2.5} levels, modeled as a continuous variable, and both overall and fatal incident MI by age at baseline (<65/≥65 y), road traffic noise L_{den} (<30.4/30.4–56.2/≥56.2), obesity (yes/no), physical activity (low/medium/high), smoking status (never/previous/current), self-reported hypertension (yes/no), self-reported diabetes (yes/no), level of urbanization (rural/provincial/urban), and self-reported menopausal status (pre-/post-menopausal), defined as cessation of bleeding, was assessed through the introduction of interaction terms into the model and tested with log-likelihood tests. In supplemental analyses, we performed the same effect modification analyses for a 1-y running mean of residential PM\textsubscript{2.5} levels.

We also conducted several other sensitivity analyses. We evaluated the effect of the following additional variables on the association between air pollution and MI: body mass index (BMI), self-reported diabetes, self-reported hypertension, and average municipality income in 1993 by further adjusting Model 2 for each of these factors independently. We also explored potential calendar effects caused by decreasing exposure levels throughout the study period. In doing so, we assessed pollutant correlations at baseline and at six different time points throughout the study period as well as mean exposure levels for each pollutant throughout the study period. We tested three different methods of adjusting for calendar effects by repeating our 1- and 3-y mean PM\textsubscript{2.5} analyses with stratifications on calendar year and birth cohort as well as with adjustment for calendar year as a linear term. As part of this, we examined correlations between each of our pollutants and both calendar year and birth cohort. We further examined the association between air pollution and fatal incident MI, restricted to deaths without prior DNPR MI diagnoses (those identified by the DRCID only), and the impact of individual confounders on the association between air pollution and incident MI by further adjusting Model 1 for each confounder independently.

Deviations from linearity for the association between exposure and outcome for our main analyses were assessed using a likelihood ratio test. Statistical analyses were conducted using Stata (version 14.2; StataCorp) and R (version 3.5.0; R Development Team). Exposure maps were created using features of the Spatial Analyst extension to Esri’s ArcMap (version 10.7.1; Esri).

Results

Of the 28,731 nurses recruited to the Danish Nurse Cohort in either 1993 or 1999, four initial exclusions were made on the basis of inactivity within the Civil Registration System prior to baseline. We excluded an additional 5,552 (19.3%) nurses with missing information on one or more covariates, potential confounders, or
Table 1. Descriptive statistics for 22,882 female nurses from the Danish Nurse Cohort at year of cohort entry in 1993 or 1999, by outcome.

| Variables | Total (n = 22,882) | MI (n = 22,241) | Overall incident MI (n = 641) | Fatal incident MI (n = 121) |
|-----------|-------------------|-----------------|-----------------------------|-----------------------------|
| Age [y [mean (SD)]] | 52.8 (7.7) | 52.6 (7.6) | 59.2 (9.2) | 65.0 (10.3) |
| Body mass index [kg/m² [mean (SD)]] | 23.7 (3.5) | 23.7 (3.5) | 24.3 (3.8) | 24.1 (3.4) |
| Body mass index [kg/m² [n (%)]] | 553 (2.4) | 535 (2.4) | 18 (2.8) | 4 (3.3) |
| Underweight (<18.5) | 15,820 (69.1) | 15,427 (69.4) | 393 (61.3) | 71 (58.7) |
| Normal weight (18.5–25) | 5,214 (22.8) | 5,033 (22.6) | 181 (28.2) | 37 (30.6) |
| Overweight (25–30) | 1,295 (5.7) | 1,246 (5.6) | 49 (7.6) | 9 (7.4) |
| Smoking status [n (%)] | 8,224 (35.9) | 8,054 (36.2) | 170 (26.5) | 28 (23.1) |
| Never | 6,877 (30.1) | 6,713 (30.2) | 164 (25.6) | 27 (22.3) |
| Previous | 7,781 (34.0) | 7,474 (33.6) | 307 (47.9) | 66 (54.5) |
| Alcohol consumption [drinks/week [n (%)]] | 3,426 (15.0) | 3,275 (14.7) | 151 (23.6) | 30 (24.8) |
| None (0) | 14,210 (62.1) | 13,852 (62.3) | 358 (55.9) | 69 (57.0) |
| Moderate (1–15) | 5,246 (22.9) | 5,114 (23.0) | 132 (20.6) | 22 (18.2) |
| High | 20,759 (90.7) | 20,195 (90.8) | 564 (80.0) | 109 (90.1) |
| Physical activity [n (%)] | 1,454 (6.4) | 1,390 (6.2) | 64 (10.0) | 26 (21.5) |
| Low | 15,277 (67.0) | 14,834 (66.3) | 439 (68.5) | 78 (64.5) |
| Medium | 6,155 (26.9) | 6,017 (27.1) | 138 (21.5) | 17 (14.0) |
| Fruit consumption [n (%)] | 828 (3.6) | 805 (3.6) | 23 (3.6) | 6 (5.0) |
| Rarely | 6,516 (28.5) | 6,324 (28.4) | 192 (30.0) | 34 (28.1) |
| Few times per week | 15,538 (67.9) | 15,112 (67.9) | 426 (65.5) | 81 (66.9) |
| Daily or several times per day | 2,123 (9.3) | 2,046 (9.2) | 77 (12.0) | 12 (9.9) |
| Avoids fatty meat [n (%)] | 20,054 (87.6) | 19,592 (88.1) | 462 (72.1) | 77 (63.6) |
| Yes | 2,828 (9.8) | 2,128 (9.6) | 108 (16.8) | 19 (15.7) |
| Hypertension [n (%)] | 22,628 (98.9) | 22,008 (99.0) | 620 (96.7) | 110 (90.9) |
| No | 254 (1.1) | 233 (1.0) | 24 (3.3) | 11 (9.1) |
| Diabetes [n (%)] | 20,054 (87.6) | 19,592 (88.1) | 462 (72.1) | 77 (63.6) |
| No | 2,236 (60.1) | 2,128 (60.7) | 108 (16.8) | 19 (15.7) |
| Oral contraceptive use [n (%)] | 11,504 (50.3) | 11,364 (51.1) | 140 (21.8) | 15 (12.4) |
| Premenopausal status | 11,378 (49.7) | 10,877 (48.9) | 501 (78.2) | 20 (16.5) |
| Postmenopausal | 11,504 (50.3) | 11,364 (51.1) | 140 (21.8) | 15 (12.4) |
| Employment status [n (%)] | 16,677 (72.9) | 16,267 (73.1) | 410 (64.0) | 82 (67.8) |
| Actively working | 2,236 (9.8) | 2,128 (9.6) | 108 (16.8) | 19 (15.7) |
| Past | 9,428 (41.2) | 9,139 (41.1) | 289 (45.1) | 57 (47.1) |
| Widowed | 2,593 (11.3) | 2,531 (11.4) | 62 (9.7) | 11 (9.1) |
| Retired | 2,123 (9.3) | 2,046 (9.2) | 77 (12.0) | 12 (9.9) |
| Single | 13,497 (60.1) | 13,192 (60.7) | 245 (38.2) | 23 (19.0) |
| Menopausal status | 11,378 (49.7) | 11,077 (48.9) | 501 (78.2) | 20 (16.5) |
| Prenopausal | 11,504 (50.3) | 11,364 (51.1) | 140 (21.8) | 15 (12.4) |
| Retired | 3,481 (15.2) | 3,374 (15.2) | 107 (16.7) | 17 (14.0) |
| Urbanization level [n (%)] | 164,487.4 (24,855.9) | 164,546.4 (24,914.5) | 162,442.1 (22,652.3) | 160,236.4 (22,208.6) |
| Rural | 9,428 (41.2) | 9,139 (41.1) | 289 (45.1) | 57 (47.1) |
| Provincial | 9,973 (43.6) | 9,728 (43.7) | 245 (38.2) | 47 (38.8) |
| Urban | 3,481 (15.2) | 3,374 (15.2) | 107 (16.7) | 17 (14.0) |
| Air pollution levels at cohort entry (1-y mean) | 164,487.4 (24,855.9) | 164,546.4 (24,914.5) | 162,442.1 (22,652.3) | 160,236.4 (22,208.6) |
| PM2.5 levels [µg/m³ [mean (SD)] | 19.6 (3.6) | 19.6 (3.6) | 21.0 (3.4) | 21.5 (3.3) |
| PM10 levels [µg/m³ [mean (SD)] | 23.5 (3.9) | 23.4 (3.9) | 25.2 (3.6) | 25.8 (3.6) |
| NO2 levels [µg/m³ [mean (SD)] | 12.6 (8.1) | 12.6 (8.0) | 13.6 (9.6) | 14.2 (10.7) |
| NOX levels [µg/m³ [mean (SD)] | 19.1 (24.4) | 19.1 (24.2) | 22.0 (29.7) | 24.4 (34.9) |
effect modifiers and 292 (1%) with missing exposure data, leaving 22,882 (79.6%) for the final main analyses. These 22,882 nurses were followed for an average of 18.6 y, during which time there were 641 cases of incident MI of which 121 were fatal (of which 104 did not have a previous DNPR MI diagnosis). Mean age at baseline was 52.8 y. Participants who developed an MI during the study period were more likely to be overweight or obese, current and heavy smokers, nondrinkers, less physically active; to have hypertension; to consume less fruit and more fatty meat; to have hypertension; to be widowed, retired, postmenopausal, never users of oral contraceptives, and hypertensive. Excluded participants were also more likely than included participants to have been current smokers, moderate alcohol consumers, less physically active, and consumers of fatty meat (see Table S1).

The correlations [measured by the Spearman’s rank correlation coefficient (ρ)] between PM_{2.5} and NO_{2}, PM_{10} and L_{den}, and NO_{2} and L_{den} were 0.65, 0.36, and 0.61, respectively (see Table S2). Correlations were stronger in urban than in provincial and rural areas. Correlations between the pollutants at six different time points between 1990 and 2014 demonstrated that with the exception of those between PM_{2.5} and PM_{10}, which decreased from 0.71 to 0.60, correlation coefficients between pollutants did not change over time (see Table S3). Additional temporal trends in exposure (mean pollutant levels) throughout the study period are shown in Figure S1. Maps depicting spatial differences in exposure to each pollutant at baseline are shown in Figure S1. Although PM_{2.5} and PM_{10} levels are highly correlated in urban areas (see Table S2), higher levels of PM_{2.5} in southeastern Denmark and of PM_{10} along the Danish west coast highlight notable source differences for these pollutants; PM_{2.5} levels are influenced by secondary pollution from central Europe, and PM_{10} levels are influenced by sea spray. NO_{2} and NO_{x}, primarily traffic related, are substantially higher in urban areas (Figure 1).

We observed positive associations between 1- and 3-y running means of PM_{2.5} and overall incident MI in our models adjusted for age and baseline year (Model 1) with a hazard ratio (HR) of 1.17 (95% CI: 1.05, 1.31) and HR = 1.20 (95% CI: 1.07, 1.35) per IQR increase of 5.3 µg/m³, respectively. These were attenuated in models further adjusted for potential confounding by lifestyle factors and CVD risk factors (Model 2) to HR = 1.05 (95% CI: 0.92, 1.19) and HR = 1.06 (95% CI: 0.92, 1.23) (Table 2). For fatal incident MI, we observed strong associations with 1- and 3-y means of PM_{2.5}, with HR = 1.75 (95% CI: 1.40, 2.18) and HR = 1.69 (95% CI: 1.33, 2.13), respectively, which, in Model 2, were attenuated to HR = 1.47 (95% CI: 1.12, 1.93) and HR = 1.35 (95% CI: 1.01, 1.81), respectively. Model 2 associations between PM_{2.5} and both overall and fatal incident MI were robust to further adjustment for road traffic noise in Model 3 (Table 2) and seemed not to deviate from linearity (Figure 2).

Similar trends were observed for PM_{10}, with 1- and 3-y mean exposure. Model 2 estimates [HRs (95% CIs)] for overall incident MI: HR = 1.04 (95% CI: 0.92, 1.17) and HR = 1.06 (95% CI: 0.93, 1.20) per IQR increase of 5.5 µg/m³, respectively. As with PM_{2.5}, stronger associations were observed among the fatal subset, with 1- and 3-y mean exposure estimates of HR = 1.43 (95% CI: 1.12, 1.81), and HR = 1.28 (95% CI: 0.98, 1.65), respectively, for Model 2. No evidence of an association was detected for NO_{2} or NO_{x}. Model 2 estimates for 1- and 3-y running means of NO_{2} were HR = 1.04 (95% CI: 0.94, 1.16) and HR = 1.05 (95% CI: 0.94, 1.16) per IQR increase of 8.1 µg/m³, whereas those for NO_{x} were HR = 1.02 (95% CI: 0.97, 1.07), and HR = 1.02 (95% CI: 0.98, 1.07) per IQR increase of 11.5 µg/m³, respectively (Table 2). Model 2 associations were robust to further adjustment for BMI, diabetes, hypertension, and average municipality income (see Table S4). However, independent assessment of each of the confounders on the Model 1 association between the pollutants and incident MI showed that employment status was most responsible for the observed attenuation in models further adjusted for potential confounding (see Table S5). Sensitivity analyses restricted to fatal MI events identified through the DRCD produced stronger estimates than analyses based on the entire subset of incident fatal events (see Table S6). Our effect estimates were also sensitive to all three methods of adjustment for calendar effects. For overall incident MI, all three models produced larger effect estimates than our original analyses. For fatal incident MI, the results were inconsistent. The models with continuous adjustment for, and stratification on, calendar year produced smaller effect sizes, whereas the model stratifying on birth cohort produced a larger effect estimate (see Table S7). Strong negative correlations were observed between calendar year and both PM_{2.5} and PM_{10}, but not with NO_{2}, NO_{x}, or L_{den}, whereas birth cohort (in 5-y strata) was not correlated with any of the exposure variables (see Table S8).

We found effect modification of the association between a 3-y running mean of PM_{2.5} and overall incident MI by diabetes, with a

| Variables | Total (n = 22,882) | MI (n = 22,241) | Overall incident MI (n = 641) | Fatal incident MI (n = 121) |
|-----------|------------------|----------------|-----------------------------|--------------------------|
| L_{den} [dB (mean (SD))] | 52.8 (8.2) | 52.8 (8.1) | 52.8 (8.6) | 53.1 (8.2) |
| L_{den} [dB (n [%])] | 7,509 (33.1) | 7,292 (33.0) | 217 (34.2) | 40 (34.5) |
| Low (<50.4) | 7,582 (33.4) | 7,379 (33.4) | 203 (32.0) | 36 (31.0) |
| Moderate (50.4–56.2) | 7,624 (33.6) | 7,410 (33.6) | 214 (33.8) | 40 (34.5) |
| High (≥56.2) | 7,509 (33.1) | 7,292 (33.0) | 217 (34.2) | 40 (34.5) |

Note: DKK, Danish Crowns; L_{den}, joint expression of day (L_d: 0700–1900 hours), evening (L_e: 1900–2200 hours), and night (L_n: 2200–0700 hours) road traffic noise levels, with 5 and 10 dB penalties added to L_d and L_n, respectively; MI, myocardial infarction; NO_{2}, nitrogen dioxide; NO_{x}, nitrogen oxides; PM, particulate matter; PM_{2.5}, PM with an aerodynamic diameter of <2.5 µm/m³; PM_{10}, PM with an aerodynamic diameter of <10 µm/m³; SD, standard deviation.

Self-reported (diagnosed/medication for).

In 1993.

Among 21,403 participants with available air pollution data at the year of cohort entry.

Among 22,715 participants with available road traffic noise data at the year of cohort entry.

Annual weighted (5 dB penalty added to L_d and 10 dB penalty added to L_n) mean.
strong positive association among those with diabetes [HR = 2.15 (95% CI: 1.08, 4.25)] and no association in those without [HR = 1.02 (95% CI: 0.88, 1.18)] (p interaction = 0.04) (Table 3).

Effect modification analyses using a 1-y running mean of PM$_{2.5}$ were very similar to those based on the 3-y running mean (see Table S9).

Table 2. Associations between long-term exposure to air pollution (per IQR exposure increment) and both overall and fatal incident MI in the Danish Nurse Cohort.

| Pollutants | Overall incident MI | Fatal incident MI |
|------------|---------------------|-------------------|
|            | Model 1$^a$ | Model 2$^b$ | Model 3$^c$ | Model 1$^a$ | Model 2$^b$ | Model 3$^c$ |
| n (cases)  | HR (95% CI) | HR (95% CI) | HR (95% CI) | HR (95% CI) | HR (95% CI) | HR (95% CI) |
| PM$_{2.5}$ | 22,884 (641) | 22,884 (641) | 22,878 (641) | 22,955 (123) | 22,955 (123) | 22,878 (121) |
| PM$_{10}$  | 1.17 (1.05, 1.31) | 1.05 (0.92, 1.19) | 1.04 (0.91, 1.19) | 1.75 (1.40, 2.18) | 1.47 (1.12, 1.93) | 1.48 (1.13, 1.95) |
| NO$_2$     | 1.03 (0.95, 1.13) | 1.04 (0.94, 1.16) | 1.04 (0.93, 1.17) | 1.04 (0.86, 1.26) | 1.05 (0.83, 1.31) | 1.08 (0.85, 1.38) |
| NO$_x$     | 1.03 (0.98, 1.07) | 1.02 (0.97, 1.07) | 1.02 (0.97, 1.08) | 1.03 (0.93, 1.14) | 1.02 (0.91, 1.13) | 1.03 (0.92, 1.15) |

| n (cases)  | HR (95% CI) | HR (95% CI) | HR (95% CI) | HR (95% CI) | HR (95% CI) | HR (95% CI) |
| PM$_{2.5}$ | 22,887 (641) | 22,887 (641) | 22,882 (641) | 22,887 (121) | 22,887 (121) | 22,882 (121) |
| PM$_{10}$  | 1.20 (1.07, 1.35) | 1.06 (0.92, 1.23) | 1.06 (0.92, 1.23) | 1.69 (1.33, 2.13) | 1.35 (1.01, 1.81) | 1.38 (1.03, 1.85) |
| NO$_2$     | 1.04 (0.95, 1.13) | 1.05 (0.94, 1.16) | 1.05 (0.94, 1.17) | 1.02 (0.84, 1.23) | 1.01 (0.80, 1.28) | 1.05 (0.81, 1.34) |
| NO$_x$     | 1.03 (0.98, 1.07) | 1.02 (0.98, 1.07) | 1.02 (0.97, 1.08) | 1.02 (0.92, 1.13) | 1.01 (0.90, 1.12) | 1.02 (0.91, 1.14) |

| n (cases)  | HR (95% CI) | HR (95% CI) | HR (95% CI) | HR (95% CI) | HR (95% CI) | HR (95% CI) |
| PM$_{2.5}$ | 22,902 (642) | 22,902 (642) | 22,902 (642) | 22,902 (121) | 22,902 (121) | 22,902 (121) |
| NO$_2$     | 1.04 (0.96, 1.13) | 1.10 (0.99, 1.22) | 1.13 (1.01, 1.27) | 1.01 (0.84, 1.22) | 1.08 (0.86, 1.36) | 1.12 (0.87, 1.43) |
| NO$_x$     | 1.03 (1.00, 1.07) | 1.04 (1.00, 1.08) | 1.05 (1.00, 1.09) | 1.02 (0.94, 1.11) | 1.03 (0.94, 1.12) | 1.04 (0.94, 1.14) |

Note: Estimates are based on IQR increments for PM$_{2.5}$ (<5.3 μg/m$^3$), PM$_{10}$ (<5.5 μg/m$^3$), NO$_2$ (<8.1 μg/m$^3$), and NO$_x$ (<11.5 μg/m$^3$), and L$_{den}$ (10 dB). CI, confidence interval; HR, hazard ratio; IQR, interquartile range; L$_{den}$, joint expression of day (L$_{d}$; 0700–1900 hours), evening (L$_{e}$; 1900–2200 hours), and night (L$_{n}$; 2200–0700 hours) road traffic noise levels, with 5 and 10 dB penalties added to L$_{e}$ and L$_{n}$, respectively; MI, myocardial infarction; NO$_2$, nitrogen dioxide; NO$_x$, nitrogen oxides; PM, particulate matter; PM$_{2.5}$, PM with an aerodynamic diameter of <2.5 μg/m$^3$; PM$_{10}$, PM with a diameter <10 μg/m$^3$.

$^a$Model 1: adjusted for age (underlying time) and year of cohort entry (1993 or 1999).

$^b$Model 2: Model 1 further adjusted for smoking status, pack-years, consumption of fruit, avoidance of fatty meat, alcohol consumption, use of oral contraceptives, use of hormone therapy, physical activity, marital status, employment status, and level of urbanization.

$^c$Model 3: Model 2 further adjusted for L$_{den}$.

Figure 1. Smoothed maps of annual residential exposure levels for (A) PM$_{2.5}$, (B) PM$_{10}$, (C) NO$_2$, and (D) NO$_x$ in the Danish Nurse Cohort at year of cohort entry (1993 or 1999). Note: NO$_2$, nitrogen dioxide; NO$_x$, nitrogen oxides; PM, particulate matter; PM$_{2.5}$, PM with a diameter <2.5 μg/m$^3$; PM$_{10}$, PM with a diameter <10 μg/m$^3$. 

Table S9.
observed stronger associations between NO₂ and fatal incident MI, detected stronger associations between PM₂.₅ and fatal cardiovascular events (Brook et al. 2010). In brief, they include our findings among the subset of fatal cases.

We detected strong associations of fatal incident MI with PM₂.₅ and PM₁₀ but none with NO₂ or NOₓ. These results are similar to those of Rosenlund et al. (2009, 2006) who reported stronger associations between PM₁₀ and fatal MI than with all incident MI. In another Danish prospective cohort study, Roswall et al. (2017) observed stronger associations between NO₂ and fatal incident MI, as compared with all incident MI, and other studies have also reported stronger associations with fatal, compared with nonfatal, cardiovascular outcomes. Puett et al. (2008) detected enhanced associations between PM₁₀ and fatal coronary heart disease (CHD) as compared with overall incident CHD, and Miller et al. (2007) detected stronger associations between PM₂.₅ and cardiovascular mortality than with PM₂.₅ and first cardiovascular event. Our findings are also in line with a large body of evidence supporting links between air pollution and cardiovascular mortality (Brook et al. 2010; Hoek et al. 2002; Lelieveld et al. 2019). Together, these findings support the premise that air pollution may be more strongly associated with fatal cardiovascular events. This could be because air pollution is associated with more severe cardiovascular events, fewer MIs that can be treated upon admission, or more out-of-hospital deaths. However, limitations in the quality of cause-of-death data for out-of-hospital sudden MI deaths among people with no prior MI diagnoses warrant caution in the interpretation of our findings among the subset of fatal cases.

There are a number of known and suspected biological mechanisms involved in the pathogenesis of air pollution–induced cardiovascular events (Brook et al. 2010). In brief, they include direct effects on both the pulmonary and cardiovascular systems, as well as indirect effects, all of which operate through chronic and systemic inflammatory pathways. Recently it has been suggested that air pollution, acting through inflammatory pathways, may promote cardiovascular events, such as MIs, by triggering vulnerable plaque and increasing thrombogenicity (Xu et al. 2019). Because exposure to both short- (Xu et al. 2019) and long- (Lucht et al. 2019) term air pollution has been associated with markers of such inflammation, including C-reactive protein (Lucht et al. 2019; Xu et al. 2019) and fibrinogen (Xu et al. 2019), it is difficult to discern the exact timescales on which these mechanisms operate.

Animal and human studies suggest the progression of atherosclerosis (a condition promoted by inflammation) as a main mechanism behind air pollution–induced MI and CHD development (Brook et al. 2010; Sun et al. 2005). Associations between air pollution and subclinical atherosclerosis support this (Kaufman et al. 2016; Künzli et al. 2010; Sun et al. 2005). Although the lack of observed effects for any pollutant on overall MI incidence in multiple large epidemiological studies is noteworthy (Atkinson et al. 2013; Bodin et al. 2016; Lipsett et al. 2011; Miller et al. 2007; Puett et al. 2011, 2009, 2008; Rosenlund et al. 2006; Stockfelt et al. 2015), recent evidence of an association between those particles capable of entering the bloodstream, PM₂.₅ (Bai et al. 2019a; Danesh Yazdi et al. 2019; Klompmaker et al. 2019; Yusuf et al. 2020) and UFPs (Bai et al. 2019b; Downward et al. 2018), and incident MI supports this mechanistic evidence. This may imply that air pollution–promoted atherosclerosis leads to a more severe subtype of ischemic event that results in death. Stronger associations have generally been observed with stroke than with MI (Miller et al. 2007) and other coronary events (Hoffmann et al. 2015), which supports this. Alternatively, advancements in early diagnosis and treatment may result in more angioplasties, which subsequently prevent MI. Interestingly, Miller et al. (2007) found that PM₂.₅ was positively associated with coronary revascularizations. Rather than air pollution causing more severe cardiovascular events, another
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Table 3. Effect modification of the association between residential PM$_{2.5}$ concentration (modeled as a continuous variable, 3-y mean, per IQR increase) and both overall and fatal incident MI in the Danish Nurse Cohort.

| Variables                          | Overall incident MI | Fatal incident MI |
|-----------------------------------|---------------------|-------------------|
|                                   | n       | Model 3 [HR (95% CI)]$^a$ | p-Value$^b$ | n       | Model 3 [HR (95% CI)]$^a$ | p-Value$^b$ |
| Age at baseline (y)               | n       |                       |              | n       |                       |              |
| <65                               | 472     | 1.00 (0.82, 1.20)     |              | 61      | 1.15 (0.68, 1.94)     |              |
| ≥65                               | 169     | 1.17 (0.89, 1.53)     | 0.50        | 60      | 1.38 (0.92, 2.08)     | 0.71        |
| Road traffic noise [dB (L$_{den}$)]$^c$ | 178     | 1.08 (0.81, 1.42)     |              | 36      | 1.95 (1.11, 3.41)     |              |
| Low (<50.4)                       | 211     | 1.03 (0.78, 1.37)     |              | 38      | 1.25 (0.66, 2.35)     |              |
| Medium (50.4–56.2)                | 252     | 1.07 (0.87, 1.33)     | 0.70        | 47      | 1.28 (0.83, 1.99)     | 0.80        |
| High (≥56.2)                      | 592     | 1.06 (0.92, 1.23)     |              | 112     | 1.37 (1.01, 1.86)     |              |
| Obesity$^d$                        | 49      | 0.87 (0.49, 1.52)     | 0.57        | 9       | 1.68 (0.42, 6.75)     | 0.87        |
| Physical activity                 | n       |                       |              | n       |                       |              |
| Low                               | 64      | 1.01 (0.68, 1.50)     |              | 26      | 1.46 (0.87, 2.45)     |              |
| Medium                            | 439     | 1.05 (0.88, 1.26)     |              | 78      | 1.27 (0.85, 1.89)     |              |
| High                              | 138     | 1.10 (0.78, 1.55)     | 0.71        | 17      | 1.61 (0.63, 4.07)     | 0.73        |
| Smoking status                    | n       |                       |              | n       |                       |              |
| Never                             | 170     | 0.92 (0.69, 1.23)     |              | 28      | 1.20 (0.66, 2.16)     |              |
| Previous                          | 164     | 1.25 (0.95, 1.66)     |              | 27      | 2.18 (1.19, 4.00)     |              |
| Current                           | 307     | 1.02 (0.83, 1.26)     | 0.63        | 66      | 1.16 (0.76, 1.77)     | 0.21        |
| Hypertension$^e$                  | n       |                       |              | n       |                       |              |
| No                                | 462     | 1.01 (0.85, 1.20)     |              | 77      | 1.24 (0.85, 1.81)     |              |
| Yes                               | 179     | 1.09 (0.84, 1.43)     | 0.26        | 44      | 1.56 (0.95, 2.55)     | 1.00        |
| Diabetes$^e$                      | n       |                       |              | n       |                       |              |
| No                                | 620     | 1.02 (0.88, 1.18)     |              | 110     | 1.25 (0.90, 1.73)     |              |
| Yes                               | 21      | 2.15 (1.08, 4.25)     | 0.04        | 11      | 2.63 (0.76, 9.07)     | 0.27        |
| Level of urbanization             | n       |                       |              | n       |                       |              |
| Rural                             | 289     | 0.92 (0.73, 1.15)     |              | 57      | 1.33 (0.82, 2.14)     |              |
| Provincial                        | 245     | 1.16 (0.91, 1.47)     |              | 47      | 1.30 (0.82, 2.08)     |              |
| Urban                             | 107     | 1.21 (0.90, 1.62)     | 0.60        | 17      | 1.48 (0.77, 2.77)     | 0.67        |
| Menopausal status                 | n       |                       |              | n       |                       |              |
| Premenopausal                     | 140     | 1.01 (0.69, 1.48)     |              | 15      | 1.63 (0.58, 4.58)     |              |
| Postmenopausal                    | 501     | 1.03 (0.88, 1.21)     | 0.25        | 106     | 1.32 (0.97, 1.80)     | 0.48        |

Note: Estimates are based on IQR increments of 5.3 μg/m$^3$. CI, confidence interval; HR, hazard ratio; IQR, interquartile range; L$_{den}$, joint expression of day (L$_d$; 0700–1900 hours), evening (L$_e$; 1900–2200 hours), and night (L$_n$; 2200–0700 hours) road traffic noise levels, with 5 and 10 dB penalties added to L$_d$ and L$_e$, respectively; MI, myocardial infarction; PM$_{2.5}$, particulate matter with an aerodynamic diameter of <2.5 μg/m$^3$.

$^a$Model 3 was adjusted for age (underlying time), year of cohort entry (1993 or 1999), smoking status, year of birth, and 10-y running means. Although we used the same exposure estimation model in our own study, we did not observe associations between NO$_2$ for any exposure window investigated, among either overall or fatal MI. Despite these differences, the large estimate discrepancies observed between this same limita-

$^b$From likelihood ratio test for interaction.

$^c$Body mass index ≥30.

$^d$Self-reported (diagnosed/medication for).

possible explanation could be that the relatively weak association between these is now masked by better secondary or tertiary prevention and early therapy. This could be, in part, due to the fact that MI has more clinical precursors, such as angina, which facilitate earlier hospital contacts. Air pollution also likely acts through multiple mechanisms and may exacerbate preexisting cardiovascular conditions to contribute to the development of cardiovascular events, fatal or otherwise (Brook et al. 2004; Mann et al. 2002). However, we did not find effect modification of the association between a 3-y mean exposure to PM$_{2.5}$ and either overall or fatal incident MI by self-reported hypertension, which may be considered a proxy and/or precursor for other cardiovascular conditions. Nurses with hypertension or other cardiovascular conditions may seek out medical monitoring and/or treatment that ultimately prevents MI, or at least prevents severe, fatal MI, possibly explaining this. We did find effect modification of this association by self-reported diabetes, which is in line with a large body of epidemiological evidence suggesting associations between exposure to air pollution and diabetes (Bulti et al. 2014), especially among women (Eze et al. 2015; Shin et al. 2019), and supported by experimental evidence of biological links between systemic inflammation and impaired glucose metabolism (Sun et al. 2009). Although the single validation study of MI in the DRCD suggests that MI as an underlying or contributory cause of death is reliable (Madsen et al. 2003), this validation was predominantly based on cases for which hospital records of an MI exist, as opposed to cases for which autopsy results were available. The autopsy rate in Denmark has dropped from 75% (of all in-hospital deaths) in the 1970s, to <10% by 2011 (Helweg-Larsen 2011), suggesting a decline in the quality of data in the DRCD that should not be overlooked. It is also likely that the stronger associations among fatal cases of incident MI observed in the two Swedish case–control studies (Rosenlund et al. 2009, 2006) suffer from this same limitation. In the Danish Diet, Cancer, and Health cohort, Roswall et al. (2017) observed associations between a 5-y running mean exposure of NO$_2$ and both incident [HR = 1.07 (95% CI: 1.02, 1.11) per 6.8 μg/m$^3$] and fatal incident [HR = 1.15 (95% CI: 1.06, 1.26) per 7.0 μg/m$^3$] MI and similar findings for exposure windows of 1- and 10-y running means. Although we used the same exposure estimation model in our own study, we did not observe associations for NO$_2$ for any exposure window investigated, among either outcome. There are notable differences between the study by Roswall et al. (2017) and ours, including the populations sampled (urban vs. nationwide) and the defining criteria for cases of fatal incident MI. Despite these differences, the large estimate discrepancies observed between two Danish studies is stark and warrants further prospective cohort investigations into the differential effects of air pollution on severity of MI in other populations. Of the recent studies on long-term exposure to PM$_{2.5}$ and incident MI, three prospective cohort studies (Bai et al. 2019a;
Danesh Yazdi et al. 2019; Yusuf et al. 2020) observed an association, whereas one (Downward et al. 2018) did not. Among 5.1 million long-term residents of Ontario, a 3-y running mean exposure to PM$_{2.5}$ was associated with hospital-admission based acute MI incidence [HR = 1.05 (95% CI: 1.04, 1.06), per 3.5 μg/m$^3$] (Bai et al. 2019a). A 3-y running mean exposure to PM$_{2.5}$ was also associated with MI incidence [HR = 1.03 (95% CI: 1.00, 1.06), per 10 μg/m$^3$] in the Prospective Urban and Rural Epidemiology cohort, based on 21 countries and 155,722 participants (Yusuf et al. 2020). Among Medicare participants in seven southeastern states in the United States, each 1-μg/m$^3$ increase in annual mean PM$_{2.5}$ concentration was associated with first hospital MI admission [HR = 1.02 (95% CI: 1.01, 1.02)] (Danesh Yazdi et al. 2019). Our PM$_{2.5}$ estimate for the 3-y running mean [HR = 1.06 (95% CI: 0.92, 1.23) per 5.3 μg/m$^3$] is of a similar effect size to the 3-y mean estimates reported by Bai et al. (2019a) and Yusuf et al. (2020), and we believe our estimates are more in line with the results from these studies than those from Downward et al. (2018), who reported a nonsignificant, although inverse, association between PM$_{2.5}$ and incident MI [HR = 0.83 (95% CI: 0.44, 1.57) per 5 μg/m$^3$] among the Dutch branch of the European Prospective Investigation into Cancer and Nutrition study (Downward et al. 2018). However, Downward et al. (2018) also reported associations with both PM$_{10}$ coarse (PM with an aerodynamic diameter of 2.5–10 μm) and UFPs (Downward et al. 2018), supporting the need for further investigation into different size fractions of PM.

Our findings also corroborate those from earlier large cohort and case–control studies (Lipsett et al. 2011; Madrigano et al. 2013; Miller et al. 2007; Puetz et al. 2011, 2009). Among 65,893 women in the Women’s Health Initiative cohort, Miller et al. (2007) detected associations between long-term exposure to PM$_{2.5}$ and incidence of overall cardiovascular disease as well as CHD, cerebrovascular disease, stroke, and coronary revascularization but not MI incidence [HR = 1.06 (95% CI: 0.85, 1.34) per 10 μg/m$^3$]. In the California Teacher’s Study (124,614 women), Lipsett et al. (2011) found no association between MI incidence and PM$_{2.5}$ [HR = 0.98 (95% CI: 0.83, 1.16) per 10 μg/m$^3$] nor PM$_{10}$, NO$_2$, or NO$_x$. In line with our findings, Madrigano et al. (2013) detected associations with area-level PM$_{2.5}$, but none with total PM$_{2.5}$, in a case–control study based on 4,467 MI cases from the Worcester Heart Attack Study [odds ratio (OR) = 1.04 (95% CI: 0.96, 1.11) per 1.05 μg/m$^3$]. The majority of studies on long-term exposure to PM$_{2.5}$, including this one, suggest that the evidence for an association with incident MI is weak. However, given that a number of recent studies that may benefit from improved exposure modeling methods, have found evidence for an association, and given that the suggestion of an association with fatal incident MI, there is still a need to further explore these relationships. Because PM$_{10}$ is partially comprised of PM$_{2.5}$ and because these two pollutants are moderately correlated throughout our study period (see Table S3), we cannot fully separate the effects of each of these pollutants. However, it remains useful to compare our PM$_{10}$ estimates with those from other studies. In line with the majority of studies on PM$_{10}$ (Atkinson et al. 2013; Downward et al. 2018; Klompmaker et al. 2019; Lipsett et al. 2011; Puetz et al. 2011, 2008; Rosenlund et al. 2006), we did not observe evidence of an association with incident MI. For an association between PM$_{10}$ and incident MI, Downward et al. (2018) reported an HR = 1.27 (95% CI: 0.77, 2.09) per 10 μg/m$^3$. Atkinson et al. (2013) reported an HR = 1.01 (95% CI: 0.98, 1.05) per 3 μg/m$^3$ among 836,557 British general practice patients. Puetz et al. (2008) reported an HR = 1.09 (95% CI: 0.92, 1.29) per 10 μg/m$^3$ among 66,250 nurses from the American Nurses’ Health Study, and in a later study on 17,545 male health professionals in the Health Professionals Follow-Up Study, Puett et al. (2011) reported an HR = 1.05 (95% CI: 0.92, 1.22) per 7 μg/m$^3$. Further, Rosenlund et al. (2009, 2006) reported ORs (95% CI) of 1.00 (0.79, 1.27) and 1.04 (1.00, 1.09) per 5 μg/m$^3$ in two Swedish case–control studies with 1,397 and 24,347 MI cases, respectively.

Finally, we observed no evidence of an association between MI incidence and NO$_2$ or NO$_x$. These findings corroborate those of the majority of studies that have explored long-term exposure to either of these pollutants (Atkinson et al. 2013; Bai et al. 2019a, 2019b; Bodin et al. 2016; Downward et al. 2018; Lipsett et al. 2011; Puett et al. 2008; Rosenlund et al. 2009, 2006; Roswall et al. 2017; Stockfelt et al. 2015), although they are inconsistent with two that detected significant positive associations with NO$_2$ (Grazuleviciene et al. 2004; Roswall et al. 2017). To date, no study has observed an association between long-term exposure to NO$_2$ and incident MI. Of the studies on NO$_2$, three in particular (Bai et al. 2019a, 2019b; Roswall et al. 2017) serve as useful comparators owing to similarities in exposure estimation methods and statistical modeling techniques. For exposure to 3-y running means, Bai et al. (2019a, 2019b) reported HR = 1.01 (95% CI: 0.99, 1.02) per 13.9 ppb (Bai et al. 2019a) and HR = 1.01 (95% CI: 0.99, 1.03) per 4.0 ppb (Bai et al. 2019b) among residents throughout the entirety of Ontario and its capital city Toronto, respectively. The effect sizes of these estimates are similar to that in our Model 2 of a 3-y running mean exposure [HR = 1.05 (95% CI: 0.94, 1.16) per 8.1 μg/m$^3$]. As previously discussed, our study has a number of similarities to the study by Roswall et al. (2017) of the Danish Diet, Cancer and Health cohort, including the air pollution estimation methods, the use of running exposure means, and case indentification methods (for overall incident MI), and it is unclear why Roswall et al. (2017) observed associations with incident and fatal incident MI for 1-, 5-, and 10-y running means of NO$_2$, whereas we did not. As a major source of NO$_2$ exposure is road traffic, it is possible that Roswall et al. (2017) were better able to detect associations with incident MI among the urban Danish Diet, Cancer and Health cohort. The only other study to report an association between long-term exposure to NO$_2$ and incident MI was also based on an urban population (Grazuleviciene et al. 2004). Although all other studies have reported null associations, only one has been based on an urban population (Bai et al. 2019b), and it would be interesting to explore the effects of NO$_2$ and NO$_x$ further among populations that may be prone to higher exposures. Our sensitivity analyses exploring the effects of BMI, self-reported diabetes and hypertension, and average municipality income (Models 4–7; see Table S4) did not deviate substantially from our Models 2 or 3 (Table 2), which were further adjusted for exposure to road traffic noise. Knowledge of the effect of noise exposure on the association between air pollution and incident MI is limited: Although our results support one recent study that found that the association between UFPs and MI was unchanged following adjustment for noise (Bai et al. 2019b), they are less consistent with a large Swiss study that found substantial attenuation of the association between air pollution and MI mortality after adjusting for noise (Héritier et al. 2019). Moreover, we did not see any difference in effect estimates following adjustment for municipality-level income (see Table S4). However, we cannot exclude the possibility of residual confounding by unmeasured spatial factors related to one’s neighborhood, such as green space or special emissions sources, which may be uniquely tied to socioeconomic status, and our inability to adjust for these factors is a notable limitation.

The question of which timescale is best to use for Cox models in time-varying air pollution studies and how best to account for calendar effects remains unresolved. Our sensitivity analyses
testing different methods of controlling for these effects produced HRs for associations with PM$_{2.5}$ that were substantially larger than those in our main analyses, although confidence intervals between these and our main estimates were overlapping (see Table S7). In their performance assessment of Cox models with different timescales used to study time-varying environmental exposures, Griffin et al. (2012) reported that in cases where the environmental exposure was highly correlated with calendar time, models performed worse when calendar time was either used as the underlying timescale or adjusted for as a covariate when using age or time on study as the underlying scale. Given strong inverse correlations between both PM$_{2.5}$ and PM$_{10}$ and calendar year during our study period (see Table S8), we believe the associations from the models adjusting for, and stratifying on, calendar year, are likely an artifact of collinearity. Furthermore, Griffin et al. (2012) noted that models fit with age as the underlying timescale, which did not adjust for calendar time, remained relatively robust. Therefore, we do not believe that secular trends in our exposures are likely to be causing a bias in our original models, despite the lack of adjustment for calendar effects. It has also been suggested that for outcomes in which the effect of age on one’s hazard is strong, the best method of controlling for calendar effects is through the use of age as the underlying timescale with stratification on birth cohort (Korn et al. 1997). In our models stratified on birth cohort, we saw an increase in estimate size relative to our original models (see Table S7) although correlation between all our pollutants and birth cohort was low (see Table S8). Because our original estimates were the lowest of any of models adjusting for calendar effects, we believe it is unlikely that the lack of adjustment for calendar effects has induced an association in these models. However, given these findings, the impact of different methods of calendar-effect adjustment on effect size should be explored in other cohorts.

The strengths of this study primarily stem from its prospective cohort design, large size, and unique ability to link various population registries together for high-quality exposure and outcome assessments. Our study also benefits from state-of-the-art exposure modeling methods and information on residential noise exposure—an important, yet often unadjusted for, potential confounder of the association between air pollution and MI risk. We did not have data on out-of-hospital MI deaths for the first half of our study period because we did not have access to the DRCD after 31 December 2013. Further, we did not have data on revascularizations to construct a combined outcome. Given the observed special importance of air pollution on fatal MI outcomes, we hypothesize that this may have diluted the effect sizes of the estimates for fatal MIs.

Because of the missing exposure and covariate data, we excluded a substantial portion of the Danish Nurse Cohort in this study. Although we cannot exclude the possibility of selection bias, we do not have reason to believe that missingness on covariates or exposure was dependent on the outcome. Potential misclassification, both in outcome and exposure, are also limitations of this study. We were unable to obtain true personal exposures that accounted for commute or workplace air pollution levels and possible overlooked confounding by sources of indoor air pollution. Further, the likelihood of error in the exposure estimates during our study should not be ignored, and validations of our model in 1999 and 2009 suggest that it may have been less precise during the first half of our study (Hvidtfeldt et al. 2018). Given the relatively small values of the IQR exposure increments for PM$_{2.5}$ (5.3 $\mu$g/m$^3$) and PM$_{10}$ (5.5 $\mu$g/m$^3$) in this study, even small errors in exposure estimates could have significant implications for exposure misclassification. However, we believe that potential misclassification related to both air pollution modeling errors and the measurement of confounders will be random and tend to move the results toward the null hypothesis. Exposure misclassification was equally inevitable for residential traffic noise given that information on personal exposure modifiers such as nighttime earplug usage or insulation by window or other building materials could not be accounted for.

**Conclusion**

Evidence on long-term exposure to air pollution and incident MI is inconsistent, especially in light of the unknown contribution of road traffic noise. This prospective cohort study of 22,882 Danish female nurses >44 years of age is the first to examine the association between PM and incident MI adjusting for road traffic noise. We did not observe associations between PM$_{2.5}$, PM$_{10}$, NO$_2$, or NO$_x$ with overall incident MI. However, we observed associations between PM$_{2.5}$ and PM$_{10}$ among a subset of fatal incident MI cases. Our findings were robust to adjustment for road traffic noise; however, further investigation is warranted.

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