Comparison of transient associations of air pollution and AMI hospitalisation in two cities of Alberta, Canada, using a case-crossover design

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ABSTRACT

Objective: To investigate reproducibility of outcomes for short-term associations between ambient air pollutants and acute myocardial infarction (AMI) hospitalisation in 2 urban populations.

Design: Using a time-stratified design, we conducted independent case-crossover studies of AMI hospitalisation events over the period 1999–2010 in the geographically close and demographically similar cities of Calgary and Edmonton, Alberta, Canada. Patients with his/her first AMI hospitalisation event were linked with air pollution data from the National Ambient Pollution Surveillance database and meteorological data from the National Climatic Data Center database. Patients were further divided into subgroups to examine adjusted pollution effects.

Results: Effects of pollution levels with 0–3-day lag were modelled using conditional logistic regression and adjusted for daily average ambient temperature, dew point temperature and wind speed.

Setting: Population-based studies in Calgary/Edmonton.

Participants: 12 066/10 562 first-time AMI hospitalisations in Calgary/Edmonton.

Main outcome measures: Association (adjusted OR) between daily ambient air pollution levels and hospitalisation for AMI.

Conclusions: Comparison of independent investigations of the effect of air pollution on risk of AMI hospitalisation in Calgary and Edmonton, Alberta, indicated that none of the effect variables were reproduced in the 2 cities, despite geographic closeness (within 300 km of each other), and demographic and air pollution similarities.

INTRODUCTION

Numerous epidemiological studies have described evidence of adverse associations between air pollution and hospital admission or emergency room visits for myocardial infarction (MI).1–21 This included a recent systematic review that reported significant associations with MI for all air pollutants except ozone (O3).2 While the associations are to some extent plausible, mechanisms underlying these associations are not fully understood.19 In addition, concerns persist about the modelling approach, covariate selection and other confounders that can lead to very different results.22 23 To highlight this, we summarised literature from PubMed related to case-crossover studies of relationships between particulate matter (PM) air pollution and MI published before 15 March 2015. Nineteen studies1–21 of PM and MI with greater than 1000 MI events were identified and are listed in table 1. From table 1, it can be observed that study findings do not always agree with each other, even for studies with very large numbers of observations.3 5 19

Another feature of the studies in table 1 was the difference in location of populations studied, including populations in cities of USA3 7–9 14 17 19 20 Europe,4–6 11–13 21 Australia/New Zealand19 and Taiwan.10 15 16 Our interest was in understanding whether...
### Table 1: Review of case-crossover studies in literature for association between PM and MI

| Study            | Location                      | Participants | Exposure        | Design          | Findings                                                                 |
|------------------|-------------------------------|--------------|-----------------|-----------------|--------------------------------------------------------------------------|
| Talbott et al.   | Washington DC and 4 east      | 688 715 cases of MI | PM$_{2.5}$      | Time-stratified | No association for lag 0 and 1 day with acute MI in 2 east coast states for all seasons |
|                  | coast US states               |              |                 |                 |                                                                          |
| Wichmann et al.  | Gothenburg, Sweden            | 28 215 cases of MI | PM$_{10}$, soot | Time-stratified | No association found                                                      |
| Milojevic et al. | England and Wales             | 452 343 cases of MI | PM$_{2.5}$, PM$_{10}$ | Time-stratified | No association found                                                      |
| Bard et al.      | Strasbourg, France            | 2134 cases of MI | PM$_{10}$       | Time-stratified | No association found                                                      |
| Hodas et al.     | New Jersey, USA               | 1561 HA for transmural MI (age ≥18) | PM$_{2.5}$       | Time-stratified | Refined ambient PM$_{2.5}$ (24 h average before onset) was associated with transmural MI |
| Rich et al.      | New Jersey, USA               | 1562 HA for transmural MI (age ≥18) | PM$_{2.5}$       | Time-stratified | PM$_{2.5}$ species (24 h average before onset) was associated with transmural MI |
| Kioumourtzoglou et al. | 3 US cities                   | Emergency HA | OC species      | Modified bidirectional | No association found                                                  |
| Tsai et al.      | Taipei, Taiwan                | 27 563 HA for acute MI | PM$_{10}$       | Time-stratified | PM$_{10}$ (1–6 h average before onset) was associated with acute MI |
| Bhaskaran et al. | England and Wales             | 79 288 HA for MI | PM$_{10}$       | Time-stratified | PM$_{10}$ (lag 2 day) was associated with acute MI |
| Nuvolone et al.  | Florence, Italy               | 11 450 HA for acute MI | PM$_{10}$       | Time-stratified | PM$_{10}$ was associated with acute MI |
| Cadum et al.     | 10 Italian cities             | HA for acute MI | PM$_{10}$       | Time-stratified | PM$_{2.5}$ (24 h average before onset) was associated with transmural MI |
| Rich et al.      | New Jersey, USA               | 5864 HA for first-time AMI | PM$_{2.5}$       | Time-stratified | PM$_{10}$ (3-day average before onset) was associated with MI |
| Hsieh et al.     | Taipei, Taiwan                | 23 420 HA for MI | PM$_{10}$       | Time-stratified | PM$_{10}$ (3-day average before onset) in cool days (<25°C) was associated with MI |
| Cheng et al.     | Kaohsiung, Taiwan             | 9349 HA for MI | PM$_{10}$       | Time-stratified | PM$_{10}$ (lag 0-day) was associated with acute MI |
| Zanobetti and Schwartz | Boston, USA                  | 15 578 HA for acute MI | PM$_{2.5}$, BC  | Time-stratified | PM$_{2.5}$ (24 h average before onset) was associated with MI |
| Barnett et al.   | 5 cities in Australian and New Zealand | HA for CVD (age ≥15) | PM$_{2.5}$, PM$_{10}$ | Time-stratified | PM$_{10}$ (lag 0–2 days) was associated with acute MI |
| Zanobetti and Schwartz | 21 US cities                 | 302 453 HA for MI (age ≥65) | PM$_{10}$       | Time-stratified | PM$_{10}$ (lag 0–2 days) was associated with acute MI |
| Sullivan et al.  | Washington, USA               | 5793 cases of acute MI | PM$_{2.5}$, PM$_{10}$ | Time-stratified | No association found                                                                 |
| D’Ippoliti et al. | Rome, Italy                   | 6531 HA for first-time AMI | TSP            | Time-stratified |                                                                                                     |

Here we focus only on studies of association between PM and MI, they could be partial results from larger studies. Only studies with >1000 MI events are reported.

AMI, acute myocardial infarction; BC, black carbon; HA, hospital admission; MI, myocardial infarction; OC, organic carbon; PM, particulate matter; TSP, total suspended particulate.
population characteristics might play a role in influencing outcomes for these types of studies. To explore this further, we hypothesised that two demographically similar Canadian cities with similar large populations, climate (weather) and air pollution characteristics should exhibit comparable (reproducible) air pollution effects for MI. We undertook and compared results for independent case-crossover studies in the two main cities of Alberta (Calgary and Edmonton) to test this hypothesis.

Calgary (~1.1 million people, elevation 1045 m above sea level, latitude 51° 2′ N, longitude 114° 3′ W) and Edmonton (~820 000 people, elevation 645 m above sea level, latitude 53° 32′ N, longitude 113° 29′ W) are geographically close to each other (<300 km apart) and both located on the east side of Rockies in western Canada. The Calgary-Edmonton population centres and corridor in between is the most urbanised area in Alberta. Both cities have a relatively moderate semiarid climate with warm summers and cool winters. Both can be subject to wide variation in weather patterns; for example, temperature below −35°C in winter and above 35°C in summer. Both cities share similarities in air pollution characteristics (described later), population structure (age distributions), as well as in some important risk factors for acute myocardial infarction (AMI) disease (table 2). There are also dissimilarities between the cities, such as average prevalence rates of smoking, diabetes and obesity. Our objective was to demonstrate that air pollution effects for MI were consistent between the cities in order to confirm their reliability beyond traditional statistical significance (ie, p<0.5). Being able to independently reproduce results in the two cities leads to more realistic effects that represent our target population of interest (urban Alberta), not just Calgary or Edmonton.

**MATERIALS AND METHODS**

**Data source**

From the provincial ministry of health, we requested all historical records of hospital admission for AMI (International Classification of Diseases (ICD) 10 code I21-I22 or ICD-9 code 410) for Calgary and Edmonton urban dwellers, respectively, for the study. We received de-identified records with a unique scrambled ID of each patient (first name, surname, date of birth) and temporarily linked them to a data file. These records were further averaged across the four stations. The time series of daily average values of the five pollutants were calculated from hourly concentration and further averaged across the four stations. The time series of daily average concentrations of the five pollutants were linked with AMI hospitalisation data for each of the two cities. We did not consider sulfur dioxide (SO2) in the analysis because of lack of data. SO2 is primarily monitored at stations close to industrial activities which, for the most part, are located away from where the populations are in Alberta.

Daily meteorological data during the study period were obtained from the US National Climatic Data Center (NCDC). For four stations in the metropolitan area of Calgary (NCDC ID 712350, 713930, 718778 and 718770) and for four stations in metropolitan area of Edmonton (NCDC ID 711210, 711570, 713510 and 718790) provide historical daily meteorological records for air temperature (daily average, minimum and maximum temperature in °C), daily average dew point temperature (in °C) and daily average wind speed (in knots). These records were further averaged across the four stations to represent daily average levels of temperature, dew point temperature and wind speed in each city. These time-series data were linked with AMI data for each of the two cities.
Study design and analysis

The case-crossover design was used to study each city separately. The case-crossover design was developed from the case-control design to study associations of transient exposures with acute events. An investigator samples only cases with this design and compares each transient exposure with acute events. An investigator from the case

selection of the reference periods. The whole study period was stratified into calendar months, and all days in the same year, same month and matching weekday of the hazard exposure day were selected as controls. A time-stratified reference-selection design is reported as a preferred approach for minimising referents that are not chosen a priori and are functions of the observed event times (referred to as overlap bias). A conditional logistic regression model was fitted and statistical parameters (coefficient, p value, OR and lower/upper bounds of 95% CI) were calculated for each of the cities and each of 600 effect variables, defined by five cohort or subcohorts (main, male, female, agecat1 (age <65), agecat2 (age ≥65)), six subgroups (whole, STEM, NSTEMI, hypertension, diabetes, dysrhythmia), five pollutants (CO, NO, NO2, O3, PM2.5) and 4 lag times (0–3 days). Each of the models was adjusted with three metrological variables (daily average of temperature, dew point temperature and wind speed). A stepwise selection procedure was adopted to

Table 2 Demographic information and important risk factors of acute myocardial infarction (AMI) for Edmonton and Calgary populations

| AMI risk factor             | Prevalence in Calgary | Prevalence in Edmonton |
|----------------------------|----------------------|------------------------|
|                            | Both     | Female | Male    | Both     | Female | Male    |
| Smoking                    | 14.89    | 12.75  | 16.99   | 18.13    | 14.68  | 21.57   |
| Hypertension               | 8.66     | 8.59   | 8.76    | 8.77     | 8.65   | 8.92    |
| Diabetes                   | 4.01     | 3.56   | 4.52    | 4.78     | 4.42   | 5.19    |
| Obesity                    | 15.07    | 13.20  | 16.77   | 17.16    | 15.22  | 18.93   |
| History of coronary heart disease | 2.40 | 1.79  | 3.06    | 2.34     | 1.72   | 3.02    |
| Age 0–19 years             | 0.25     | 0.25   | 0.26    | 0.25     | 0.25   | 0.26    |
| Age 20–64 years            | 0.62     | 0.61   | 0.62    | 0.60     | 0.59   | 0.61    |
| Age ≥65 years              | 0.13     | 0.14   | 0.12    | 0.15     | 0.16   | 0.13    |
| Unemployment               | 4.10     | 4.40   | 3.90    | 4.70     | 4.70   | 4.80    |

Unemployment data for the two cities were from Census 2006 (http://www12.statcan.gc.ca/census-recensement/2006/index-eng.cfm); all other data were from Alberta Interactive Health Data Application (http://www.ahw.gov.ab.ca/IHDA_Retrieval/) and calculated from annual prevalence rates over the period 2000–2010. Prevalence of smoking is the rate of current daily smokers; prevalence rate of obesity is the rate of people with body mass index ≥30; unemployment is the rate of unemployment for those aged 15 years or over.

Table 3 First-time hospitalisations for acute myocardial infarction in different subgroups

| City      | Cohort | Whole     | STEMI     | NSTEMI    | HTN       | Diabetes | Dysrhythmia |
|-----------|--------|-----------|-----------|-----------|-----------|----------|-------------|
| Calgary   | Main   | 12 066 (100%) | 4206 (34.9%) | 4834 (40.1%) | 6060 (50.2%) | 2844 (23.6%) | 2127 (17.6%) |
|           | Male   | 8191 (67.9%) | 3009 (24.9%) | 3106 (25.7%) | 3846 (31.9%) | 1858 (15.4%) | 1413 (11.7%) |
|           | Female | 3875 (32.1%) | 1197 (9.9%) | 1728 (14.3%) | 2214 (18.3%) | 986 (8.2%) | 714 (5.9%) |
|           | Agecat1 | 5330 (44.2%) | 2210 (18.3%) | 1804 (15.0%) | 2240 (18.6%) | 1068 (8.9%) | 585 (4.8%) |
|           | Agecat2 | 6736 (55.8%) | 1996 (16.5%) | 3030 (25.1%) | 3820 (31.7%) | 1776 (14.7%) | 1542 (12.8%) |
| Edmonton  | Main   | 10 562 (100%) | 3492 (33.1%) | 4754 (45.0%) | 6154 (58.3%) | 2825 (26.7%) | 1935 (18.3%) |
|           | Male   | 6991 (66.2%) | 2446 (23.2%) | 3008 (25.8%) | 3772 (35.7%) | 1773 (16.8%) | 1201 (11.4%) |
|           | Female | 3571 (33.8%) | 1046 (9.9%) | 1746 (16.5%) | 2382 (22.6%) | 1052 (10.0%) | 734 (6.9%) |
|           | Agecat1 | 4613 (43.7%) | 1813 (17.2%) | 2386 (21.4%) | 3892 (36.8%) | 1786 (16.9%) | 1486 (14.1%) |
|           | Agecat2 | 5949 (56.3%) | 1679 (15.9%) | 2964 (28.1%) | 3892 (36.8%) | 1786 (16.9%) | 1486 (14.1%) |

Frequency of STEMI and NSTEMI was based on the period 1 April 2002 to 31 March 2010; frequency of other subgroups was based on the period 1 April 1999 to 31 March 2010. Percentages=number of patients in subgroup divided by 12 066 (for Calgary) or 10 562 (for Edmonton).
eliminate redundant meteorological variables with critical level for variable entry and critical level for variable stay in the model both set at 0.25. Coefficient estimation and OR estimation were calculated for the IQR difference (between the 25th and the 75th centiles) for the covariate of interest. For example, for female hypertension patients, we built a logistic regression model on a subset of the data (for Edmonton or Calgary) that included all female hypertension records in each cohort when checking whether 3-day lag daily average PM$_{2.5}$ level was associated with AMI hospitalisation. The model included one variable for 3-day lag PM$_{2.5}$ level and three variables for 3-day lag meteorological condition.

RESULTS
Descriptive analysis
There were a total of 12 066 (10 562) first-time AMI hospitalisation events in the urban areas of Calgary and Edmonton over the period 1 April 1999 to March 2010—an average of 2.62 (3.00) hospitalisations per day. The number of hospitalisations for predefined subgroups is listed in table 3.

Figure 1 shows a summary of monthly average concentrations of the five air pollutants and the three climate factors over the 1 April 1999 to 31 March 2010 period in the two cities. Monthly average air pollution concentrations were not widely divergent among the cities. Obvious seasonal trends are apparent for several of the air pollutants. Much higher (lower) NO and NO$_2$ levels occur during winter (summer) which is opposite to that of O$_3$, which has lower (higher) levels occurring during winter (summer). The highest monthly PM$_{2.5}$ levels occur during the summer period (mid-June to mid-September). Overall, figure 1 did not indicate any major differences in monthly average pollution levels and trends between Calgary and Edmonton.

Estimated effects of the pollutants
The same analysis procedure—time-stratified case-crossover design and conditional logistic regression—was repeated for each of the cities and each of the 600 effect variables. For each model, we focused on reporting the estimated effect of an air pollution variable, which was adjusted with the three meteorological factors (daily mean temperature, dew point temperature and wind speed). Parameter estimates for all 600 effect variables (including coefficient, SD, p value, and OR and 95% CI) for Calgary and Edmonton are saved in the online supplementary file 2. Variables with estimated p value less than 0.05 are summarised in table 4. There were 7 (4) effect variables that exhibited significant associations (p<0.05) in Calgary (Edmonton) from 600 effect variables examined (table 4). If results for only a single city (Calgary or Edmonton) are reported, most of the variables in table 4 for that city could be suggested as exhibiting positive associations with AMI hospitalisations. However, comparing the findings from each city allows us to consider the issue of reproducibility of effects.

As stated previously, both cities share similarities in air pollution levels (figure 1) as well as in some important risk factors for AMI disease (table 2) and similar air pollution effects for MI would be anticipated for each city.

Table 4 illustrates that all of the effect variables exhibiting positive associations with AMI hospitalisations are not reproduced for each city; significant effect variables identified in Calgary were not reproduced in Edmonton, and vice versa. For example, NO$_2$ was suggested as a risk factor for several subgroups identified in Calgary, whereas no positive effect was found for NO$_2$ in any of the subgroups identified in Edmonton in table 4. Likewise, PM$_{2.5}$ was suggested to be a risk factor for several Edmonton subgroups, whereas no positive effect was found for PM$_{2.5}$ in any of the subgroups identified in Calgary.

DISCUSSION
Among the 600 potential effect variables investigated for the study in Calgary (Edmonton), we found that only 1.17% (0.67%) was statistically significant by using the traditional 5% criterion. None of the associations was reproduced in the two cities. A previous time-series analysis of emergency department visits for angina/MI at 14 hospitals in seven Canadian cities, including Edmonton, during the 1990s and early 2000s examined associations with CO, NO$_2$, O$_3$ and PM$_{2.5}$.

The strongest associations with increased emergency department visits for angina/MI were only related to 24 h average concentrations of CO and NO$_2$ lag 0 days in Edmonton, but not for any of the other Canadian cities studied (Halifax, Montreal, Ottawa, Saint John, Toronto and Vancouver).

Our study observed increased hospital visits for AMI with several subgroups in Calgary associated with 24 h average concentrations of CO and NO$_2$ lag 1 day, but not in Edmonton (table 4). As stated previously, different modelling approaches, covariate selection and other confounders and different data sets can lead to dissimilar results.

Lack of reproducibility of a PM$_{2.5}$ pollution effect on AMI in our study is not completely unexpected. Although numerous studies have reported PM$_{2.5}$ as an important risk factor for MI, including a recent systematic review concluding that most air pollutants were associated with increased short-term risk of MI, an earlier review stated that less than half of literature studies showed clear evidence of elevated MI risk from exposure to air pollutants.

In light of this, we believe that being able to reproduce findings from independent investigations employing similar methods is a useful feature for exploring air pollution effects. It is worthwhile to examine possible reasons for differences in most of the findings for these two cities. First, we speculate that there are differences in population characteristics at the individual level that the analysis did not account for—such as an omitted risk.
factor or a difference in air pollution exposure, susceptibility and/or response—in the two cities. If this is true, we should seek out these differences and further investigate air pollution-health associations at the individual level separately for each city. Meta-analysis would be unreasonable because of a specific effect instead of a random effect among the two cities. The data on population characteristics (table 2), monthly average air pollution characteristics (figure 1) and air pollutant IQR concentrations used in the analysis for each city (table 4) are not widely divergent such that a specific difference(s) among the two cities might be an explanation.

Second, differences in the findings of table 4 may be attributed to a weak association between air pollution and AMI hospitalisation in each city and/or a false finding. If this is true, larger scale data sets would be needed to reveal these associations with sufficient power. From a practical point of view, we should be ignoring weak associations. If we had to depend on a large number of health events (eg, over 300,000 AMI events) to demonstrate a weak association between air pollutants and a health outcome, the findings would be less meaningful in public health practice.

The study was an exploratory analysis comparing independent investigations of air pollution effects on risk of AMI hospitalisation in two geographically close and demographically similar cities of Alberta, Canada. It was assumed that both cities had large enough populations to satisfy epidemiological design criteria. We emphasised reproducibility of findings in the investigations as a way to explore air pollution effects on risk of hospitalisation of urban populations in Alberta. This approach, in our view, was a simple way to identify associations between air pollution and short-term health outcomes in these urban populations. The study was limited in that it was an ecological study with the exposure variables

![Figure 1](http://bmjopen.bmj.com/)

Figure 1  Seasonal trends of monthly average concentrations of air pollutants and monthly levels of climate factors (April 1999–March 2010). Left (right) column represents Calgary (Edmonton); while the top (bottom) row represents pollution (climate) levels. Unit of the monthly average concentrations of pollutants was adjusted: CO (1 unit=1 mg/m³); NO (1 unit=10 µg/m³); NO₂ (1 unit=10 µg/m³); O₃ (1 unit=10 µg/m³); PM₂.₅, fine particulate matter (1 unit=10 µg/m³). Unit of the monthly average values of climate factors: TEMP, temperature (1 unit=1°C); DEWP, dew point temperature (1 unit=1°C); WDSP, wind speed (1 unit=0.1 knots).
(air pollutants and meteorological variables) measured at central monitoring locations, and thus they did not represent actual exposures for patients with AMI. In addition, we only considered effects from CO, NO, NO₂, O₃, and PM₂.₅. Because of data limitations we could not consider other potentially important factors such as SO₂, other factors (e.g., alcohol consumption, physical activity), exposure location prior to onset of AMI (e.g., outdoor vs indoor) or special drug usage prior to onset of AMI.

**SUMMARY**

Comparison of independent investigations of air pollution effects on risk of AMI hospitalisation in Calgary and Edmonton, Alberta, indicated that none of the pollutants investigated—including CO, NO, NO₂, O₃ and PM₂.₅—showed consistent positive associations with increased risk of AMI hospitalisation. The methodology used here is proposed as a way to explore reproducibility of air pollution effects on risk of hospitalisation of urban populations in Alberta.

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**Contributors** XW participated in the design of the study, contributed in the acquisition of air pollution and meteorological data, performed the statistical analysis, and helped to draft the manuscript. WK conceived the study, participated in its design and coordination, and helped to draft the manuscript. PK contributed in the acquisition of health data. All authors read the draft paper, provided critical comments and approved the final version of the paper.

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**Competing interests** None declared.

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**Data sharing statement** Air pollution data are freely available and can be accessed from the Environment Canada National Air Pollution Surveillance (NAPS) database. Daily meteorological data can be obtained from the United States National Climatic Data Center (NCDC). The health administrative data can be obtained with an ethics approval by contacting one of the co-authors who now resides with an agency that provides anonymised administrative data can be obtained with an ethics approval by contacting one of the co-authors who now resides with an agency that provides anonymised health data. To access this data, one must have a protocol approved by an ethics board.

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**Table 4** Parameter and OR estimates for multivariate logistic regression models

| Cohort   | Subgroup | N    | Variable | Lag | Estimate | SE  | p Value | OR   | Lower CL | Upper CL |
|----------|----------|------|----------|-----|----------|-----|---------|------|----------|----------|
| Calgary  | Main     | 12066| NO₂      | 1   | 0.0452   | 0.0194 | 0.0199 | 1.046 | 1.007    | 1.087    |
|          | Female   | 3875 | NO₂      | 1   | 0.0709   | 0.0342 | 0.0381 | 1.073 | 1.004    | 1.148    |
|          | Female   | 1728 | PM₂.₅    | 0   | -0.0627  | 0.0319 | 0.0489 | 0.939 | 0.882    | 1.000    |
|          | Agecat1  | 585  | PM₂.₅    | 1   | -0.1285  | 0.0589 | 0.0292 | 0.879 | 0.784    | 0.987    |
|          | Agecat2  | 6736 | CO       | 1   | 0.0264   | 0.0130 | 0.0428 | 1.027 | 1.001    | 1.053    |
|          | Agecat2  | 6736 | NO       | 1   | 0.0324   | 0.0139 | 0.0193 | 1.033 | 1.005    | 1.061    |
|          | Agecat2  | 6736 | NO₂      | 1   | 0.0734   | 0.0260 | 0.0047 | 1.076 | 1.023    | 1.132    |
| Edmonton | Main     | 2825 | PM₂.₅    | 3   | 0.0532   | 0.0247 | 0.0314 | 1.055 | 1.005    | 1.107    |
|          | Main     | 1935 | PM₂.₅    | 3   | -0.0616  | 0.0312 | 0.0192 | 1.031 | 0.940    | 0.885    |
|          | Agecat1  | 4613 | PM₂.₅    | 1   | 0.0397   | 0.0192 | 0.0391 | 1.040 | 1.002    | 1.080    |
|          | Agecat1  | 1039 | PM₂.₅    | 1   | 0.0836   | 0.0423 | 0.0485 | 1.087 | 1.001    | 1.181    |

Data were calculated for an IQR increase of CO (0.27 mg/m³), NO (26.3 µg/m³), NO₂ (21.9 µg/m³), O₃ (26.7 µg/m³), PM₂.₅ (7.1 µg/m³) in the Calgary study, or of CO (0.27 mg/m³), NO (20 µg/m³), NO₂ (23.8 µg/m³), O₃ (30 µg/m³), PM₂.₅ (7.8 µg/m³) in the Edmonton study. Frequency of NSTEMI was based on the period 1 April 2002 to 31 March 2010; frequency of other subgroups was based on the period 1 April 1999 to 31 March 2010.

Agecat1, age <65; Agecat2, age ≥65; CL, 95% confidence level; N, number of first-time hospitalisations for acute myocardial infarction; NSTEMI, non-ST segment elevation myocardial infarction; PM, particulate matter.
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