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

Xiaoming Wang,¹ Warren Kindzierski,¹ Padma Kaul²

To cite: Wang X, Kindzierski W, Kaul P. Comparison of transient associations of air pollution and AMI hospitalisation in two cities of Alberta, Canada, using a case-crossover design. *BMJ Open* 2015;5:e009169. doi:10.1136/bmjopen-2015-009169

► Prepublication history and additional material is available. To view please visit the journal (<http://dx.doi.org/10.1136/bmjopen-2015-009169>).

Received 22 June 2015
Revised 29 August 2015
Accepted 20 October 2015

CrossMark

¹School of Public Health, University of Alberta, Edmonton, Alberta, Canada

²Department of Medicine, Canadian Vigour Centre, University of Alberta, Edmonton, Alberta, Canada

Correspondence to
Dr Warren Kindzierski;
warrenk@ualberta.ca

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. 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.

Results: Among 600 potential air pollution effect variables investigated for the Calgary (Edmonton) population, only 1.17% (0.67%) was statistically significant by using the traditional 5% criterion. None of the effect variables were reproduced in the 2 cities, despite their geographic closeness (within 300 km of each other), and demographic and air pollution similarities.

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 air pollutants investigated—CO, NO, NO₂, O₃ and particulate matter (PM_{2.5})—showed consistent positive associations with increased risk of AMI hospitalisation.

INTRODUCTION

Numerous epidemiological studies have described evidence of adverse associations

Strengths and limitations of this study

- We considered reproducibility of air pollution effects on risk of acute myocardial infarction (AMI) hospitalisation in two Alberta cities.
- We separately investigated a variety of air pollutants (CO, NO, NO₂, O₃ and particulate matter (PM_{2.5})) in each city.
- We did not include SO₂ because of data limitations.
- We focused on reproducibility of findings as a step in identifying important associations between air pollution and AMI hospitalisation in the major population centres of Alberta.

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 studies^{3–21} 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 USA,^{3–7–9 14 17 19 20} Europe,^{4–6 11–13 21} Australia/New Zealand¹⁸ 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 <i>et al</i> ⁸	Washington DC and 4 east coast US states	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
Wichmann <i>et al</i> ⁴	Gothenburg, Sweden	28 215 cases of MI	PM ₁₀ , soot	Time-stratified	No association found
Milojevic <i>et al</i> ⁵	England and Wales	452 343 cases of MI	PM _{2.5} , PM ₁₀	Time-stratified	No association found
Bard <i>et al</i> ⁶	Strasbourg, France	2134 cases of MI	PM ₁₀	Time-stratified	No association found
Hodas <i>et al</i> ⁷	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 <i>et al</i> ⁸	New Jersey, USA	1562 HA for transmural MI (age ≥18)	PM _{2.5} species	Time-stratified	PM _{2.5} species (24 h average before onset) was associated with transmural MI
Kioumourtoglou <i>et al</i> ⁹	3 US cities	Emergency HA	OC species	Modified bidirectional	No association found
Tsai <i>et al</i> ¹⁰	Taipei, Taiwan	27 563 HA for acute MI	PM ₁₀	Time-stratified	No association found
Bhaskaran <i>et al</i> ¹¹	England and Wales	79 288 HA for MI	PM ₁₀	Time-stratified	PM ₁₀ (1–6 h average before onset) was associated with acute MI
Nuvolone <i>et al</i> ¹²	Florence, Italy	11 450 HA for acute MI	PM ₁₀	Time-stratified	PM ₁₀ (lag 2 day) was associated with acute MI
Cadum <i>et al</i> ¹³	10 Italian cities	HA for acute MI	PM ₁₀	Time-stratified	PM ₁₀ was associated with acute MI
Rich <i>et al</i> ¹⁴	New Jersey, USA	5864 HA for first-time AMI	PM _{2.5}	Time-stratified	PM _{2.5} (24 h average before onset) was associated with transmural MI
Hsieh <i>et al</i> ¹⁵	Taipei, Taiwan	23 420 HA for MI	PM ₁₀	Time-stratified	PM ₁₀ (3-day average before onset) was associated with MI
Cheng <i>et al</i> ¹⁶	Kaohsiung, Taiwan	9349 HA for MI	PM ₁₀	Time-stratified	PM ₁₀ (3-day average before onset) in cool days (<25°C) was associated with MI
Zanobetti and Schwartz ¹⁷	Boston, USA	15 578 HA for acute MI	PM _{2.5} , BC	Time-stratified	PM _{2.5} (lag 0-day) was associated with acute MI
Barnett <i>et al</i> ¹⁸	5 cities in Australian and New Zealand	HA for CVD (age ≥15)	PM _{2.5} , PM ₁₀	Time-stratified	PM _{2.5} (24 h average before onset) was associated with MI
Zanobetti and Schwartz ¹⁹	21 US cities	302 453 HA for MI (age ≥65)	PM ₁₀	Time-stratified	PM ₁₀ (lag 0-day) was associated with MI
Sullivan <i>et al</i> ²⁰	Washington, USA	5793 cases of acute MI	PM _{2.5} , PM ₁₀	Time-stratified	No association found
D'Ippoliti <i>et al</i> ²¹	Rome, Italy	6531 HA for first-time AMI	TSP	Time stratified	TSP (lag 0–2 days) was associated with acute MI

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 (~1.10 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 Rock Mountains 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 establish whether 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 first-time hospitalisation events for AMI occurring during 1 April 1999 to 31 March 2010 with patients aged 20 or over and living in the urban areas of both Calgary and Edmonton. Forward sortation areas (FSAs) are geographical regions in Canada in which all postal codes start with the same three characters. Patients in urban Calgary and Edmonton eligible for the study were from FSAs shown in online supplementary file 1.

We classified a patient's event with an AMI code as I214 (ICD-10) or 4107 (ICD-9) as having non-ST segment elevation myocardial infarction (NSTEMI), and the others as ST segment elevation myocardial infarction

(STEMI). The validity of NSTEMI/STEMI classification in the early years of the data set was uncertain. Because of this we only considered first-time hospitalisation events for AMI occurring during 1 April 2002 to 31 March 2010 for classification of NSTEMI/STEMI. Secondary diagnosis codes (diagnosis 2–25) were used to define comorbidities for each patient, including hypertension (ICD-10 codes I10-I13 and I15, or ICD-9 codes 401), diabetes (ICD-10 codes E10-E14, or ICD-9 codes 250) and dysrhythmia (ICD-10 codes I47-I49, or ICD-9 codes 427). First-time hospitalisation events for AMI occurring during 1 April 1999 to 31 March 2010 were used for classification of these comorbidities.

Sex and age (at the start date of an AMI hospitalisation event) were used to define four subcohorts (male, female, patients age <65, patients age ≥65). Patients in the main cohort or in one of the four subcohorts were further divided into subgroups defined by AMI type or comorbidities, including all patients in the cohort or subcohort, patients with NSTEMI, patients with STEMI, patients with diabetes, patients with hypertension and patients with dysrhythmia.

Air pollution time-series data for Alberta were obtained from the Environment Canada National Ambient Pollution Surveillance (NAPS) database²⁴ for the period 1 March 1999 to 30 April 2010. Four stations in the urban area of Calgary (NAPS ID 90218, 90222, 90228 and 90302) and four stations in urban area of Edmonton (NAPS ID 90120, 90121, 90122 and 90130) provide hourly records of five criteria air pollutants—that is, carbon monoxide (CO), nitric oxide (NO), nitrogen dioxide (NO₂), O₃ and particulate matter with an aerodynamic diameter ≤2.5 µm (PM_{2.5}). Daily average levels of the five pollutants were calculated from hourly concentration and further averaged across the four stations. The time series of daily average levels 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).²⁵ Four stations in the metropolitan area of Calgary (NCDC ID 712350, 713930, 718770 and 718778) and 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.

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.

Study design and analysis

The case-crossover design²⁶ 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 patient's exposure during a short time period just before a case event (hazard period) with the participant's exposure at other times (reference periods) without leading to case event.²⁶ Because there is almost perfect matching of all measured and unmeasured individual characteristics that do not vary or that vary slowly over time (ie, age, gender, education, social economic status, personal lifestyle characteristics, body mass index, comorbidity conditions, etc), this design intrinsically adjusts for these characteristics.

The k-th day (k ranging from 0 to 3) before onset of an AMI hospital admission was selected as the hazard exposure period for a patient in the cohort. A time-stratified reference-selection design^{27 28} was used for

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).²⁹

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, PM_{2.5}) 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 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 (28.5%)	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%)	1790 (16.9%)	2262 (21.4%)	1039 (9.8%)	449 (4.3%)
	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).

Agecat1, age <65; Agecat2, age ≥65; HTN, hypertension; NSTEMI, non-ST segment elevation myocardial infarction; STEMI, ST segment elevation myocardial infarction.

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 NO₂ levels occur during winter (summer) which is opposite to that of O₃, 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.^{30 31} 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, NO₂ was suggested as a risk factor for several subgroups identified in Calgary, whereas no positive effect was found for NO₂ 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, NO₂, O₃ and PM_{2.5}.³² The strongest associations with increased emergency department visits for angina/MI were only related to 24 h average concentrations of CO and NO₂ 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 NO₂ lag 1 day, but not in Edmonton ([table 4](#)). As stated previously, different modelling approaches, covariate selection and other confounders^{22 23} 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,^{7 8 14 17 18} including a recent systematic review concluding that most air pollutants were associated with increased short-term risk of MI,² an earlier review stated that less than half of literature studies showed clear evidence of elevated MI risk from exposure to air pollutants.¹

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

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); PM_{2.5}, 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).

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 association³³ between air pollution and AMI hospitalisation in each city and/or a false finding.³⁴ 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^{3 5 19}) 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

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	Whole	12 066	NO ₂	1	0.0452	0.0194	0.0199	1.046	1.007	1.087
Female	Whole	3875	NO ₂	1	0.0709	0.0342	0.0381	1.073	1.004	1.148
Female	NSTEMI	1728	PM _{2.5}	0	-0.0627	0.0319	0.0489	0.939	0.882	1.000
Agecat1	Dysrhythmia	585	PM _{2.5}	1	-0.1285	0.0589	0.0292	0.879	0.784	0.987
Agecat2	Whole	6736	CO	1	0.0264	0.0130	0.0428	1.027	1.001	1.053
Agecat2	Whole	6736	NO	1	0.0324	0.0139	0.0193	1.033	1.005	1.061
Agecat2	Whole	6736	NO ₂	1	0.0734	0.0260	0.0047	1.076	1.023	1.132
Edmonton										
Main	Diabetes	2825	PM _{2.5}	3	0.0532	0.0247	0.0314	1.055	1.005	1.107
Main	Dysrhythmia	1935	PM _{2.5}	0	-0.0616	0.0312	0.0480	0.940	0.885	0.999
Agecat1	Whole	4613	PM _{2.5}	1	0.0397	0.0192	0.0391	1.040	1.002	1.080
Agecat1	Diabetes	1039	PM _{2.5}	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), PM_{2.5} (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), PM_{2.5} (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% confident level; N, number of first-time hospitalisations for acute myocardial infarction; NSTEMI, non-ST segment elevation myocardial infarction; PM, particulate matter.

(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 PM_{2.5}. Because of data limitations we could not consider other potentially important factors such as SO₂, other factors (eg, alcohol consumption, physical activity), exposure location prior to onset of AMI (eg, 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 PM_{2.5}—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.

Acknowledgements The authors would like to acknowledge the support of the Alberta Health Public Health Information Network project.

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.

Funding This research received no specific grant from any funding agency in the public, commercial or not-for-profit sectors.

Competing interests None declared.

Ethics approval University of Alberta Health Research Ethics Board-Health Panel (IREB Pro00010852).

Provenance and peer review Not commissioned; externally peer reviewed.

Data sharing statement Air pollution data are freely available and can be accessed from the Environment Canada National Ambient Pollution Surveillance (NAPS) database.²⁴ Daily meteorological data can be obtained from the United States National Climatic Data Center (NCDC).²⁴ 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 and de-identified health administrative data to researchers: XW, Research Innovation & Analytics, Alberta Health Services, 7235 2nd Floor, West Wing 11402 University Avenue Edmonton, Alberta, Canada T6G 2J3. Email: Xiaoming.Wang@albertahealthservices.ca; Tel: (780) 407-4679; Fax: (780) 407-8979.

Open Access This is an Open Access article distributed in accordance with the Creative Commons Attribution Non Commercial (CC BY-NC 4.0) license, which permits others to distribute, remix, adapt, build upon this work non-commercially, and license their derivative works on different terms, provided the original work is properly cited and the use is non-commercial. See: <http://creativecommons.org/licenses/by-nc/4.0/>

REFERENCES

1. Bhaskaran K, Hajat S, Haines A, *et al*. Effects of air pollution on the incidence of myocardial infarction. *Heart* 2009;95:1746–59.
2. Mustafic H, Jabre P, Caussin C, *et al*. Main air pollutants and myocardial infarction: a systematic review and meta-analysis. *JAMA* 2012;307:713–21.
3. Talbott EO, Rager JR, Benson S, *et al*. A case-crossover analysis of the impact of PM(2.5) on cardiovascular disease hospitalizations for selected CDC tracking states. *Environ Res* 2014;134:455–65.
4. Wichmann J, Sjöberg K, Tang L, *et al*. The effect of secondary inorganic aerosols, soot and the geographical origin of air mass on acute myocardial infarction hospitalisations in Gothenburg, Sweden during 1985–2010: a case-crossover study. *Environ Health* 2014;13:61.
5. Milojevic A, Wilkinson P, Armstrong B, *et al*. Short-term effects of air pollution on a range of cardiovascular events in England and Wales: case-crossover analysis of the MINAP database, hospital admissions and mortality. *Heart* 2014;100:1093–8.
6. Bard D, Kihal W, Schillinger C, *et al*. Traffic-related air pollution and the onset of myocardial infarction: disclosing benzene as a trigger? A small-area case-crossover study. *PLoS ONE* 2014;9:e100307.
7. Hodas N, Turpin BJ, Lunden MM, *et al*. Refined ambient PM_{2.5} exposure surrogates and the risk of myocardial infarction. *J Expo Sci Environ Epidemiol* 2013;23:573–80.
8. Rich DQ, Özkaynak H, Crooks J, *et al*. The triggering of myocardial infarction by fine particles is enhanced when particles are enriched in secondary species. *Environ Sci Technol* 2013;47:9414–23.

9. Kioumourtoglou MA, Zanobetti A, Schwartz JD, *et al.* The effect of primary organic particles on emergency hospital admissions among the elderly in 3 US cities. *Environ Health* 2013;12:68.
10. Tsai SS, Chen PS, Yang YH, *et al.* Air pollution and hospital admissions for myocardial infarction: are there potentially sensitive groups? *J Toxicol Environ Health A* 2012;75:242–51.
11. Bhaskaran K, Hajat S, Armstrong B, *et al.* The effects of hourly differences in air pollution on the risk of myocardial infarction: case crossover analysis of the MINAP database. *BMJ* 2011;343:d5531.
12. Nuvolone D, Balzi D, Chini M, *et al.* Short-term association between ambient air pollution and risk of hospitalization for acute myocardial infarction: results of the cardiovascular risk and air pollution in Tuscany (RISCAT) study. *Am J Epidemiol* 2011;174:63–71.
13. Cadum E, Berti G, Biggeri A, *et al.* The results of EpiAir and the national and international literature (article in Italian). *Epidemiol Prev* 2009;33(6 Suppl 1):113–19.
14. Rich DQ, Kipen HM, Zhang J, *et al.* Triggering of transmural infarctions, but not nontransmural infarctions, by ambient fine particles. *Environ Health Perspect* 2010;118:1229–34.
15. Hsieh YL, Yang YH, Wu TN, *et al.* Air pollution and hospital admissions for myocardial infarction in a subtropical city: Taipei, Taiwan. *J Toxicol Environ Health A* 2010;73:757–65.
16. Cheng MF, Tsai SS, Yang CY. Air pollution and hospital admissions for myocardial infarction in a tropical city: Kaohsiung, Taiwan. *J Toxicol Environ Health A* 2009;72:1135–40.
17. Zanobetti A, Schwartz J. Air pollution and emergency admissions in Boston, MA. *J Epidemiol Community Health* 2006;60:890–5.
18. Barnett AG, Williams GM, Schwartz J, *et al.* The effects of air pollution on hospitalizations for cardiovascular disease in elderly people in Australian and New Zealand cities. *Environ Health Perspect* 2006;114:1018–23.
19. Zanobetti A, Schwartz J. The effect of particulate air pollution on emergency admissions for myocardial infarction: a multicity case-crossover analysis. *Environ Health Perspect* 2005;113:978–82.
20. Sullivan J, Sheppard L, Schreuder A, *et al.* Relation between short-term fine-particulate matter exposure and onset of myocardial infarction. *Epidemiology* 2005;16:41–8.
21. D'Ippoliti D, Forastiere F, Ancona C, *et al.* Air pollution and myocardial infarction in Rome: a case-crossover analysis. *Epidemiology* 2003;14:528–35.
22. Koop G, McKittrick R, Tole L. Air pollution, economic activity and respiratory illness: evidence from Canadian cities, 1974–1994. *Environ Model Software* 2010;25:873–85.
23. Stylianou M, Nicolich MJ. Cumulative effects and threshold levels in air pollution mortality: data analysis of nine large US cities using the NMMAPS dataset. *Environ Pollut* 2009;157:2216–23.
24. National Air Pollution Surveillance Network, Environment Canada. <http://www.etc-cte.ec.gc.ca/NapsData/Default.aspx> (accessed Jan 2013).
25. Land-Based Station Data, National Climatic Data Center, National Oceanic and Atmospheric Administration. <http://www.ncdc.noaa.gov/data-access/land-based-station-data> (accessed Jan 2013).
26. Maclure M. The case-crossover design: a method for studying transient effects on the risk of acute events. *Am J Epidemiol* 1991;133:144–53.
27. Levy D, Lumley T, Sheppard L, *et al.* Referent selection in case-crossover analyses of acute health effects of air pollution. *Epidemiology* 2001;12:186–92.
28. Janes H, Sheppard L, Lumley T. Case-crossover analyses of air pollution exposure data—referent selection strategies and their implications for bias. *Epidemiology* 2005a;16:717–26.
29. Janes H, Sheppard L, Lumley T. Overlap bias in the case-crossover design, with application to air pollution exposures. *Stat Med* 2005;24:285–300.
30. Hernán MA, Wilcox AJ. Epidemiology, data sharing, and the challenge of scientific replication. *Epidemiology* 2009;20:167–8.
31. Peng RD, Dominici F, Zeger SL. Reproducible epidemiologic research. *Am J Epidemiol* 2006;163:783–9.
32. Stieb DM, Szyszkowicz M, Rowe BH, *et al.* Air pollution and emergency department visits for cardiac and respiratory conditions: a multi-city time-series analysis. *Environ Health* 2009;8:25.
33. Schwartz J. Particulate air pollution and mortality: a synthesis. *Public Health Rev* 1991;19:39–60.
34. Ioannidis JA. Why most published research findings are false. *PLoS Med* 2005;2:e124–30.