Introduction

The identification of respirable silica dust (Parks et al. 1999) and smoking (Sugiyama et al. 2010) as causes of rheumatoid arthritis (RA) suggests inhaled pollution as a possible factor in RA pathogenesis. Biologic mechanisms by which these exposures may lead to RA—inflammation, oxidative stress, and immune suppression—have also been correlated with ambient air pollution (Donaldson et al. 2001; Omara et al. 2000), leading to a hypothesis for the role of air pollution in the development of RA.

Traffic, a major source of air pollution, was investigated in relation to RA in the Nurses’ Health Study (NHS) cohort in the United States (Hart et al. 2009). Women who lived ≤ 50 m from a roadway were at increased risk of RA compared with those who lived ≥ 200 m away. However, there was no association with specific air pollutants, including particulate matter with an aerodynamic diameter ≤ 2.5 μm (PM2.5), nitrogen dioxide (NO2), and sulfur dioxide (SO2) in the same study population (Hart et al. 2013a). In a case–control study conducted in Sweden, Hart et al. (2013b) found small increases in RA risk associated with SO2 exposure in the 10th and 20th years before disease onset and an association with NO2 for only autoantibody-negative cases. The same group found no association with PM1.0 (≤ 10 μm in aerodynamic diameter). These initial studies provided some consistency of evidence against PM exposure as a risk factor for RA, but the results were inconsistent regarding other air pollutants.

The association of RA with traffic in the NHS remains unexplained in terms of identification of a causal agent, and Hart et al. (2013a) speculated that road proximity may be a proxy for non–air pollution exposures, such as noise or neighborhood. Noise is a biologically plausible etiologic agent whereby traffic noise may be perceived as stressful and produce an immunologic response (Prasher 2009). An important mechanism for physiologic effects from urban noise is through sleep disturbance (Zaharna and Guilleminault 2010). Stress from major life events has been linked with onset of RA in multiple studies (Herrmann et al. 2000); however, noise as a stressor has not been studied as a cause of RA.

We investigated proximity to traffic, ambient air pollution, and community noise as risk factors for RA in a large population-based study conducted in the Vancouver and Victoria regions of British Columbia (BC), Canada. In addition to providing data on traffic and air pollutants for comparison with previous studies of RA, our investigation of noise generates new information that could help explain the relationship of RA with proximity to roadways.

Methods

Study population. We conducted research for the present study within the Border Air Quality Study cohort (Gan et al. 2011, 2012a; University of British Columbia 2014), in which the study population was identified using birthdate and postal code(s) of residence in linked administrative databases from BC’s universal health insurance system (BC Ministry of Health 2006a, 2006b, 2006c). The BC Ministry of Health and the BC Vital Statistics Agency approved access to the data, and its use was facilitated by Population Data BC (https://www.popdata.bc.ca). All of the study procedures, including access to and linkage of administrative data, were covered under the ethics approval provided by the Behavioural Research Ethics Board of the University of British Columbia (certificate no. H05-80442). The study region encompassed the metropolitan centers of Vancouver and Victoria and surrounding areas in the same airshed. The cohort included 678,361 area residents who were registered with the Canadian health insurance plan, lived in the study area during the 5-year cohort definition period (1994–1998), and were 45–84 years of age at the start of the cohort follow-up period (1 January 1999).

We used a nested case–control design for this study of RA in order to evaluate risk in relation to a consistent window of exposure for each member of the study population (5 years before the diagnosis date). Cases and controls were selected from cohort members considered to be “at-risk” for developing RA at the start of the 4-year follow-up period.

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Proximity to Traffic, Ambient Air Pollution, and Community Noise in Relation to Incident Rheumatoid Arthritis

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BACKGROUND: The risk of rheumatoid arthritis (RA) has been associated with living near traffic; however, there is evidence suggesting that air pollution may not be responsible for this association. Noise, another traffic-generated exposure, has not been studied as a risk factor for RA.

OBJECTIVES: We investigated proximity to traffic, ambient air pollution, and community noise in relation to RA in the Vancouver and Victoria regions of British Columbia, Canada.

METHODS: Cases and controls were identified in a cohort of adults that was assembled using health insurance registration records. Incident RA cases from 1999 through 2002 were identified by diagnostic codes in combination with prescriptions and type of physician (e.g., rheumatologist). Controls were matched to RA cases by age and sex. Environmental exposures were assigned to each member of the study population by their residential postal code(s). We estimated relative risks using conditional logistic regression, with additional adjustment for median income at the postal code.

RESULTS: RA incidence was increased with proximity to traffic, with an odds ratio (OR) of 1.37 (95% CI: 1.11, 1.68) for residence ≤ 50 m from a highway compared with residence > 150 m away. We found no association with traffic-related exposures such as PM2.5, nitrogen oxides, or noise. Ground-level ozone, which was highest in suburban areas, was associated with an increased risk of RA (OR = 1.26; 95% CI: 1.18, 1.36 per interquartile range increase).

CONCLUSIONS: Our study confirms a previously observed association of RA risk with proximity to traffic and suggests that neither noise levels nor traffic-related air pollutants are responsible for this relationship. Additional investigation of neighborhood and individual correlates of residence near roadways may provide new insight into risk factors for RA.

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(1 January 1999 through 31 December 2002) and included 640,041 persons with no diagnosis code for RA [International Classification of Diseases, Version 9 (ICD-9); code 714 (rheumatoid arthritis and other inflammatory polyarthropathies codes 714.0–714.9); National Center for Health Statistics 1998] recorded in fee-for-service physician claims or for hospitalizations during the cohort definition period.

**Cases.** We identified RA cases by the ICD-9 code (714.0–714.9) listed for an outpatient or inpatient visit. The diagnosis date of each case (the index date) was defined by the first appearance of the relevant ICD-9 code. We used three definitions of RA:

- RA-ICD-9: three or more ICD-9 codes for RA recorded during follow-up (with two codes > 30 days apart)
- RA-prescription: two or more ICD-9 codes for RA recorded during follow-up (> 30 days apart) and one or more prescriptions (BC Ministry of Health 2012a) for a disease-modifying antirheumatic drug (DMARD) or steroids (oral or injectable, not topical) after the first ICD-9 code
- RA-specialist: two or more ICD-9 codes for RA recorded during follow-up (> 30 days apart) and one or more of these was for a visit to a physician specialist (i.e., a rheumatologist or internist).

Identification of RA patients based on two or more ICD-9 codes at least 30 days apart had 92% positive predictive value (PPV) compared with patient self-report within a large U.S. West Coast insurance company (MacLean et al. 2001). We required three or more ICD-9 codes to increase the specificity of our RA-ICD-9 definition. Our RA-prescription definition was based on validation studies in other populations that have shown information on medications in combination with ICD-9 codes (Singh et al. 2004) or self-report (Walitt et al. 2008) to further improve the specificity of case classification. We allowed either steroid or DMARD medications for the RA-prescription definition because DMARD use among RA cases in Canada was considered inappropriately low in a previous study of administrative billing data (Lacaille et al. 2005). Our RA-specialist case definition was based on the BC Ministry of Health Guidelines recommendation for early referral to a specialist when a new diagnosis of RA is suspected (BC Ministry of Health 2012b).

**Controls.** Ten controls per case were selected from the cohort remaining at risk on the index date of each RA case, matched to cases by age and sex.

**Exposure assessment.** Exposure estimates were individually assigned to each member of the study population for a 5-year period before their index date, based on the six-digit postal code of their residence(s). The size of an area covered by a residential postal code depends on the population density of the area; on average, a postal code includes 35 persons. A total of 97.5% of postal codes in the at-risk cohort’s residential histories were georeferenced to the block-face (one side of a city block) level.

**Traffic.** We used the commercially available DMTI CanMap road network to identify road locations and attributes (DMTI Spatial, Markham, Ontario, Canada). We defined “highways” as combined DMTI road categories 1 (expressways, usually four lanes) and 2 (principal highways, multilane conduits for intracity traffic), with average traffic counts for the two road types of 114,000 and 21,000 vehicles/day, respectively (Setton et al. 2005). “Major roads” were defined as DMTI road categories 3 (secondary highways, typically thoroughfares with multiple lanes and large traffic capacity) and 4 (major roads), used for shorter trips within the city), with average traffic counts of 15,000 and 18,000 vehicles/day. For each road type, we coded residential proximity in categories of ≤ 50 m or 50–150 m from the nearest road.

Members of the study population who lived in multiple residences during the exposure period were categorized according to the proximity at which they lived the longest. Proximity variables were compared with an unexposed category that included those who lived > 150 m from the road type during the exposure period.

**Air pollution.** For each member of the study population, monthly air pollutant levels were averaged over their residences(s) during the exposure period using two different approaches.

We used land-use regression (LUR) models to estimate individual residential exposures to black carbon, PM_{2.5}, NO_{2}, and nitrogen oxide (NO) in 2003 (Brauer et al. 2008; Henderson et al. 2007; Larson et al. 2009). The models were based on NO and NO_{2} measurements at 116 sites, PM_{2.5} measurements at a subset of 25 locations, and mobile monitoring of black carbon concentrations at a subset of 39 sites. We combined these measurements with geographic variables to develop high spatial resolution (10-m) LUR models. The coefficient of determination (R^2) of the models ranged from 0.52 for PM_{2.5} to 0.62 for NO. The models were used to generate a smooth spatial surface of predicted monthly concentrations for each air pollutant at each postal code during the study period. Previous work within our study region showed that the spatial pattern of pollution was adequately stable for extrapolation of an LUR model over a 7-year period (Wang et al. 2012).

A second air pollution assessment approach, pollutant concentrations measured at regulatory monitors [n = 24 monitors for PM_{10}, 26 for nitrogen oxides, 27 for ozone (O_{3}), 20 for SO_{2}, 21 for carbon monoxide (CO)] were assigned to postal codes by inverse-distance weighting (IDW) of the three closest monitors ≤ 50 km of the postal code for each day (Marshall et al. 2008). A day was set to missing if there were > 6 hr missing from the monitoring data for that day. From the daily averages, monthly average exposures were computed for each postal code area. A month was considered missing if there was a gap of > 5 consecutive days or if there were > 10 missing days for that month.

**Noise.** We used noise prediction software (Computer Aided Noise Abatement; Datakustik, Greifenberg, Germany) to estimate annual average community noise levels at each residence during the study period (Gan et al. 2012b). Because of model inputs required, noise exposure was only estimated for residences in metropolitan Vancouver. Noise exposure was based on data from 2003, including road type (Setton et al. 2005), traffic volume from a transportation planning model (EMME/2; INRO Consultants, Montreal, Quebec, Canada), railway data (e.g., type of train, velocity, and frequency), aircraft data from noise exposure forecasts produced by Vancouver International Airport Authority (Transport Canada 2005), and building heights and footprints. The annual A-weighted equivalent continuous noise level was calculated for a 10 x 10 m grid that was then averaged for each postal code area. This metric integrated noise levels during the day, evening, and night, with 5-dB(A) weighting added to evening noise and 10-dB(A) weighting added to night noise to reflect increased sensitivity of residents to noise during these periods. For each member of the study population, annual noise levels were averaged over their residence(s) during the exposure period.

**Statistical analyses.** Data analyses were conducted using SAS version 9.3 (SAS Institute Inc., Cary, NC, USA). We generated odds ratios (ORs) as estimates of relative risks (RRs) from conditional logistic regression models (Breslow and Day 1980), accounting for the age- and sex-matched design. We adjusted for potential confounding by neighborhood-level socioeconomic status (SES) using the median income level of the residential postal code from the 2001 Statistics Canada Census (Statistics Canada 2001), split by quintiles. There was no confounding by neighborhood SES in overall analyses; however, we retained this variable in all models based on evidence of weak confounding in subgroup analyses. Each road type was modeled separately, including the two proximity categories in one model with the unexposed category as the referent.
We identified 3,333 RA-ICD-9, 2,692 RA-prescription, and 1,911 RA-specialist cases among the 640,041 at-risk cohort members during 4 years of follow-up (Table 2). These numbers are consistent with the expected number of roughly 2,100 incident cases, estimated from published age-stratified incidence rates (Doran et al. 2002). There was a large overlap of individuals among the three case definitions, with 82% and 83% of RA-specialist cases also counted as RA-prescription or RA-ICD-9 cases, respectively, and 81% of RA-prescription cases also counted as RA-ICD-9 cases. Compared to the at-risk cohort, overall, cases were older at the study baseline and more likely to be female. The distribution of neighborhood SES was fairly comparable between cases and the at-risk cohort.

Residential proximity to a highway was associated with RA for each of the three case definitions (Table 3), with estimated risk increases of 37–39% for residence ≤ 50 m and 7–24% for 50–150 m, during the 5 years before diagnosis. ORs were somewhat attenuated when we excluded those who moved during the exposure period; for example, instead of a 37% increased risk, there was a 20% increased risk (95% CI: –13%, 65%) of RA-prescription diagnosis associated with continuous residence ≤ 50 m from a highway (70% of cases and 78% of controls did not move). There was no association with proximity to a major road. Risk increases with proximity to highways were observed across all subgroups of sex, age, and neighborhood SES, although not all elevations were statistically significant. For example, residence ≤ 50 m from a highway was associated with incidence of RA-prescription diagnosis in all subgroups (Figure 1), with ORs ranging from 1.19 to 1.61 and statistically significant elevations estimated for the younger-age, male, and lower-SES subgroups. Similar profiles were obtained for the other case definitions except for the RA-specialist case definition, for which the association between residence ≤ 50 m from a highway and risk of RA was seen in lower-SES neighborhoods (OR = 1.73; 95% CI: 1.23, 2.44) but not with higher SES (OR = 1.04).

ORs for the association between ambient air pollutants and RA are shown in Table 4 (results for NO and NO2 assessed by IDW did not differ importantly from the LUR results; see Supplemental Material, Table S2). The risk of developing RA was not increased with exposure to NO, NO2, black carbon, PM2.5, PM10, CO, or SO2. Conversely, many of the ORs were < 1. Analyses of exposure quintiles for these pollutants generally showed monotonic trends of decreasing risk by increasing category of exposure (see Supplemental Material, Table S3). Inverse

### Table 2. Characteristics of the study population at baseline (1999) [n (%)]

| Characteristic | At-risk cohort (n = 640,041) | RA-ICD-9 (n = 3,333) | RA-prescription (n = 2,692) | RA-specialist (n = 1,911) |
|----------------|-----------------------------|----------------------|-----------------------------|--------------------------|
| Age (years)    |                             |                      |                             |                          |
| 45–54          | 257,061 (40.2)              | 934 (28.0)           | 687 (25.5)                  | 567 (29.7)               |
| 55–64          | 162,246 (25.0)              | 820 (24.5)           | 683 (24.6)                  | 503 (26.3)               |
| 65–74          | 137,130 (21.4)              | 955 (28.7)           | 804 (29.9)                  | 534 (27.9)               |
| 75–84          | 85,604 (13.4)               | 624 (18.7)           | 539 (20.0)                  | 307 (16.1)               |
| Sex            |                             |                      |                             |                          |
| Female         | 335,062 (52.4)              | 2,218 (66.6)         | 1,854 (68.9)                | 1,307 (68.4)             |
| Male           | 304,979 (47.6)              | 1,115 (33.4)         | 838 (31.3)                  | 604 (31.6)               |
| Neighborhood SES (quintile) |               |                      |                             |                          |
| 1              |                             |                      |                             |                          |
| 2              | 126,005 (19.7)              | 732 (22.0)           | 580 (21.6)                  | 378 (19.8)               |
| 3              | 125,335 (19.6)              | 748 (22.4)           | 551 (20.5)                  | 380 (19.9)               |
| 4              | 123,928 (19.4)              | 625 (18.8)           | 510 (19.0)                  | 354 (18.5)               |
| 5              | 131,344 (20.5)              | 624 (18.7)           | 539 (20.0)                  | 404 (21.1)               |
| Missing        | 1,568 (0.2)                 | 7 (0.2)              | 6 (0.2)                     | 4 (0.2)                  |

### Table 1. Air pollution and noise exposure distributions and correlations in the at-risk cohort during the cohort definition period (1994–1998).

| Exposure          | LUR | IDW |
|-------------------|-----|-----|
| NO-LUR (μg/m³)    | 30.8 ± 11.6 | 27.8 (23.8–34.5) |
| NO2-LUR (μg/m³)   | 29.1 ± 5.2  | 28.4 (25.6–31.9) |
| Black carbon-LUR (μg/m³) | 1.3 ± 0.6 | 1.1 (0.9–1.5) |
| PM2.5-LUR (μg/m³) | 4.7 ± 2.4  | 4.3 (3.2–5.9) |
| PM10-IDW (μg/m³)  | 13.9 ± 0.6  | 13.8 (13.5–14.3) |
| CO-IDW (μg/m³)    | 27.1 ± 5.0  | 26.5 (23.7–31.2) |
| SO2-IDW (μg/m³)   | 6.5 ± 2.8   | 5.6 (4.5–8.4)   |
| Noise (dB(A))     | 63.5 ± 5.1  | 62.6 (59.9–66.8) |

Data are correlation coefficients unless otherwise indicated.
associations were quite consistent in subgroup analyses by age, sex, and neighborhood SES (e.g., Figure 2 shows RA-prescription in relation to NO$_2$).

Ground-level O$_3$ was associated with 15% and 26% increased risk of RA (for RA-ICD-9 and RA-prescription, respectively) for an increase across the IQR (Table 4), and 29% and 56% increased risk with O$_3$ level in the highest versus lowest quintile category (see Supplemental Material, Table S3). Cases increase across the IQR (Table 4), and 29% and 26% increased risk of RA (for RA-ICD-9 and RA-prescription, respectively) for an increase in proximity to traffic in the lower SES subgroup with bias adjustment for smoking.

**Discussion**

Through the use of the established population-based Border Air Quality Study cohort with state-of-the-art exposure modeling, we were able to efficiently evaluate the effects of proximity to traffic, ambient air pollution, and community noise on the risk of developing RA. Use of health records linked to postal codes in the Vancouver and Victoria, BC, metropolitan areas allowed us to identify the sizable number of cases needed for investigation of possibly subtle effects of environmental exposures on RA.

We observed an increased risk of RA with residential proximity to traffic and the largest risk increases were seen in association with road types of higher traffic volume and with greater proximity, suggesting a dose-response pattern. Our findings are similar to those from the NHS, where RA was associated with proximity to roadways (Hart et al. 2009) but not with specific traffic-related air pollutants such as NO$_2$ or PM$_{2.5}$ (Hart et al. 2013a). Another exposure from traffic—noise—was tested for the first time, and was not associated with RA in our study. The increased risk we observed with traffic could be due to confounding by risk factors for RA that may be more prevalent close to roadways, such as low SES, nonwhite race, and smoking. The health service records do not contain information on these variables; however, there was little confounding in our study due to neighborhood-level income, and sensitivity analyses suggested that smoking was unlikely to be an important confounder. Indeed, adjustment for multiple individual-level covariates in the NHS indicated little confounding of the association with proximity to roadways. Nevertheless, inconsistencies in our findings (diminished association among non-movers and association limited to lower SES neighborhoods for one case definition) suggest that other, nontraffic exposures underlie the association.

The present study and the previous two studies by Hart et al. (2013a, 2013b) did not observe any risk increase with PM$_{2.5}$ or PM$_{10}$, providing evidence against PM air

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**Table 3.** Risk of incident RA in relation to residential proximity to traffic during 5 years before diagnosis [ORs (95% CIs)].

| Exposure                  | RA-ICD-9                  | RA-prescription          | RA-specialist             |
|---------------------------|---------------------------|--------------------------|----------------------------|
|                           | Percent cases (n = 3,226) | Percent controls (n = 33,256) | OR (95% CI) | Percent cases (n = 2,686) | Percent controls (n = 26,857) | OR (95% CI) | Percent cases (n = 1,907) | Percent controls (n = 19,066) | OR (95% CI) |
| Distance from highway     |                           |                          |                            |
| > 150 m                   | 86.9                      | 89.3                     | 1.0 (referent)             | 87.0                      | 89.4                     | 1.0 (referent)             | 87.8                      | 89.0                     | 1.0 (referent)             |
| > 50–150 m                | 9.1                       | 7.8                      | 1.17 (1.03, 1.33)          | 9.1                       | 7.6                      | 1.24 (1.08, 1.43)          | 8.3                       | 8.1                      | 1.07 (0.90, 1.27)          |
| ≤ 50 m                    | 3.9                       | 2.8                      | 1.39 (1.16, 1.68)          | 4.0                       | 3.0                      | 1.37 (1.11, 1.68)          | 3.9                       | 2.9                      | 1.37 (1.07, 1.76)          |
| Distance from major road  |                           |                          |                            |
| > 150 m                   | 59.7                      | 61.2                     | 1.0 (referent)             | 60.6                      | 61.5                     | 1.0 (referent)             | 61.9                      | 61.7                     | 1.0 (referent)             |
| > 50–150 m                | 26.8                      | 25.9                     | 1.05 (0.97, 1.14)          | 25.7                      | 25.5                     | 1.02 (0.93, 1.12)          | 26.0                      | 25.4                     | 0.94 (0.81, 1.09)          |
| ≤ 50 m                    | 12.5                      | 12.8                     | 1.07 (0.96, 1.19)          | 13.7                      | 13.0                     | 1.07 (0.95, 1.20)          | 12.1                      | 12.9                     | 1.02 (0.92, 1.14)          |

Members of the study population who lived ≤ 150 m from a highway/major road during the 5-year exposure period were categorized according to proximity at which they lived the longest. ORs were adjusted for age, sex, and neighborhood SES.
Air pollution, noise, and risk of rheumatoid arthritis

We likewise did not observe increased risk with NO\textsubscript{2} or SO\textsubscript{2}, contrary to the findings of the Swedish study (Hart et al. 2013b). In fact, we observed inverse associations with certain air pollutants including NO\textsubscript{2}, SO\textsubscript{2}, and PM\textsubscript{10} (i.e., suggested protective effects) that were counter to our hypothesis. The range and variability of many air pollutants in the Vancouver region is lower than in other major cities, which may have hindered detection of risks limited to higher exposure levels. For example, the SO\textsubscript{2} IQR was larger in both the Swedish study (8 \mu g/m\textsuperscript{3}) and the NHS (14 \mu g/m\textsuperscript{3}) (Hart et al. 2013a) than in the present study (3.1 \mu g/m\textsuperscript{3}). Furthermore, it is possible that the induction period for these exposures is longer than the 5 years evaluated here, as associations with NO\textsubscript{2} and SO\textsubscript{2} in the Swedish study were strongest for exposure in the 10th year before disease onset. Discrepant results between the three studies for SO\textsubscript{2} and NO\textsubscript{2} leave an open question about the role of these pollutants in RA.

Noise has not been studied in relation to RA in humans. In rats, 90- to 95-dB noise exposure for a 1-hr period for 7 days was associated with earlier onset and increased severity of collagen-induced arthritis (Rogers et al. 1983), but there are few other existing data for evaluating the relationship. The noise metric we used represented an average community noise level across day, evening, and night hours, and ranged from < 25 to 98.5 dB(A). This metric was associated with coronary heart disease mortality in a previous investigation within the cohort (Gan et al. 2012a), with 9% increased risk per 10-dB(A) noise level that was independent of traffic-related air pollutants. We did not see any association with the risk of RA. Aside from the possibility of a true null association, it is also possible that average noise level does not capture the aspects of noise that could be etiologically relevant for RA, such as peak levels.

Ground-level O\textsubscript{3} was associated with increased risk of RA in the present study. The previous epidemiologic studies of RA have not reported on O\textsubscript{3} (Hart et al. 2013a, 2013b). O\textsubscript{3} causes its toxic effects through oxidation or peroxidation of biomolecules—directly or via free-radical reactions—potentially causing altered methylation or protein binding sites in DNA and thus generating autoantigens that may lead to autoimmunity (Kannan 2006; Kirkham et al. 2011). O\textsubscript{3} is formed in the ambient environment through the reaction of hydrocarbons with nitrogen oxides in the presence of ultraviolet light.

Because O\textsubscript{3} is itself reactive, areas with high levels of nitrogen oxides and locations close to traffic tend to have low O\textsubscript{3} levels. The areas with the highest O\textsubscript{3} levels in our study region were suburban neighborhoods with higher SES (Marshall et al. 2009). Confounding from factors spatially associated with these neighborhoods may be of concern, such as white race/ethnicity, obesity, and physical activity levels.
inactivity (Marshall et al. 2009). Nevertheless, due to limited knowledge about risk factors for RA, few variables are readily identifiable as likely confounders. Because the Border Air Quality Study was enumerated using health insurance registration records, it is a comprehensive, population-based sample that is less subject to selection biases than a study requiring active participation. Furthermore, the near-complete health coverage in Canada allowed us to identify RA cases in a population not influenced by access to health care. These strengths are unfortunately paired with weaknesses. Identification of incident RA cases from information in health service billing records is inherently imperfect, and although our case definitions were informed by validation in other populations, case diagnoses in our study were not confirmed. Our data were further limited by a lack of certain details that would be valuable for distinguishing case groups, such as autoantibody-positivity [for rheumatoid factor (RF) or anticitrullinated peptide antibodies (ACPA)]. Such details may be important because the association with proximity to roadways in the NHS was stronger for RF-positive cases (Hart et al. 2007) and the Swedish case–control study found risk increases associated with NO$_2$ for ACPA-negative cases only (Hart et al. 2013b).

The present study capitalized on extensive exposure modeling conducted by a multidisciplinary team, in which air pollution levels were assigned to individual members of the study population using methods that incorporate both intraurban spatial contrasts and temporal variability. Nevertheless, there are limitations to our models. Exposures were estimated at residential locations only, and therefore did not include air pollutant exposure inputs from workplace locations and time spent in traffic. Despite this potential misclassification, a major of the person’s time is spent at their residence (Leech et al. 2002), making ambient exposure in residential neighborhoods an important regulatory target. Another concern is that there are indoor as well as outdoor sources of the pollutants we studied. Correlations between ambient and personal exposures have been shown to be stronger for PM than for the gaseous pollutants, including O$_3$ (Sarnat et al. 2001). We understand this potential measurement error, and stress that our finding of an O$_3$ association with RA should be investigated in other study populations.

Conclusions

Our findings confirm a previously observed association of RA risk with proximity to traffic, and suggest that neither noise levels nor traffic-related air pollutants are responsible for this relationship. Additional investigation of neighborhood and individual correlates of residence near roadways may provide new insight into risk factors for RA.

References

BC Ministry of Health. 2006a. Consolidation File (BC Ministry of Health Services Registration & Premium Billing (R&P) files). Available: https://www.popdata.bc.ca/data/internal/demographic/consolidationfile [accessed 4 May 2014].

BC Ministry of Health. 2006b. Discharge Abstracts Database (Hospital Separations). Available: https://www.popdata.bc.ca/data/internal/health/dad [accessed 4 May 2014].

BC Ministry of Health. 2006c. Medical Services Plan (MSP) Paired Air pollution with a case–control study. Eur Respir J 21:936–944. doi:10.1183/09031936.06034801.

Breslow NE, Day NE. 1980. Statistical Methods in Cancer Research, Vol 1. The Analysis of Case–Control Studies. IARC Sci Publ 32:5–338.

Brunsdon C, Stone V, Tatem A, MacNee W. 2001. Ambient particle inhalation and the cardiovascular system: potential mechanisms. Environ Health Perspect 109(suppl 4):523–527.

Burkholder JD, Crouse JD, Howard MA, Anderson RE. 1998. Trends in incidence and mortality in rheumatoid arthritis in Rochester, Minnesota, over a forty-year period. Arthritis Rheum 42:625–631.

Gar G, Hawkes HW, Koechlin M, Brauer M. 2012a. Association of long-term exposure to community noise and traffic-related air pollution with coronary heart disease mortality. Am J Epidemiol 175:899–906.

Gar G, Koechlin M, Davies HW, Demers PA, Tamburic L, Brauer M. 2011. Long-term exposure to traffic-related air pollution and the risk of coronary heart disease hospitalization and mortality. Environ Health Perspect 119:501–507; doi:10.1289/ehp.1002511.

Gar G, McLean K, Brauer M, Chiarello SA, Davies HW. 2012b. Modeling population exposure to community noise and air pollution in a large metropolitan area. Environ Res 116:111–116.

Hart JE, Källberg H, Laden F, Costenbader KH, Bellander T, Costenbader KH. 2007. Traffic pollution and increased risk of rheumatoid arthritis in Rochester, Minnesota, over a forty-year period. Arthritis Rheum 51:207–211. doi:10.1002/art.22306.

Hart JE, Källberg H, Laden F, Costenbader KH. 2003. Ambient air pollution and the risk of rheumatoid arthritis in Rochester, Minnesota, over a forty-year period. Arthritis Rheum 49:1190–1196.

Hart JE, Laden F, Puett RC, Costenbader KH, Karslaw EW. 2009. Exposure to traffic pollution and increased risk of rheumatoid arthritis. Environ Health Perspect 117:1065–1069. doi:10.1289/ehp.0800503.

Henderson SB, Beckerman B, Jerrett M, Brauer M. 2007. Application of land use regression to estimate long-term concentrations of traffic-related nitrogen oxides and fine particulate matter. Environ Sci Technol 41:2422–2428.

Heermann M, Schömmerich J, Straub RH. 2000. Stress and rheumatic diseases. Rheum Dis Clin North Am 26:737–763.

Kannel WB. 2006. Free radical theory of autoimmunity. Theor Biol Med Model 3:22; doi:10.1186/1742-4682-3-22.

Kirkham PA, Caramori G, Casolari P, Papi AA, Edwards M, Herrmann M, Schölmerich J, Straub RH. 2000. Stress and rheumatoid arthritis: a population study. Arthritis Rheum 44:638–643.

Koehoorn M, Demers P, Setton EM, Hystad PW, Keller CP. 2005. Road Classification Schemes—Good Indicators of Traffic Volume? U Vic SSL Working Paper 05–014. Spatial Sciences Laboratories Occasional Papers Series 2005. Available: http://web.uvic.ca/~srss10/SSLtemp/SSLL05-014-TRAFFIC.pdf [accessed 11 November 2013].

Singh JA, Holmgren AR, Noorbaloochi S. 2004. Accuracy of Veterans Administration databases for a diagnosis of rheumatoid arthritis. Arthritis Rheum 51:952–957.

Statistics Canada. 2001. Census of Canada, Government of Canada. Available: http://www12.statcan.ca/english/ census01/home/index.cfm [accessed 4 May 2014].

Statistics Canada. 2002. Canadian Community Health Survey (CCHS) Detailed information for 2000–2001 (Cycle 1.1). Available: http://web2.statcan.gc.ca/imdb/p2s-t2p.jsp?func=能得到Survey&BVarId=1630&InstId=3359&SDDS=3226 [accessed 4 May 2014].

Sugiyama D, Nishimura K, Tamaki K, Tsuji G, Nakazawa T, Morinobu A, et al. 2010. Impact of noise as a risk factor for developing rheumatoid arthritis: a meta-analysis of observational studies. Ann Rheum Dis 69:70–81.

Transport Canada. 2005. TP1247—Aviation—Land Use in the Vicinity of Airports. Available: http://www.tc.gc.ca/eng/civilaviation/publications/tp1247-menu-1418.htm [accessed 12 May 2014].

University of British Columbia. 2014. Border Air Quality Study. Available: http://bqas.sph.ubc.ca/border-air-quality-study/ welcome-border-air-quality-study [accessed 12 May 2015].

Walit BT, Constantinescu F, Katz JD, Weinstein A, Wang H, Hernandez RK, et al. 2008. Validation of self-report of rheumatoid arthritis and systemic lupus erythematosus: the Women’s Health Initiative. J Rheumatol 35:811–818.

Wang R, Henderson SB, Shihi H, Allen RW, Brauer M. 2012. Temporal stability of land use regression models for traffic-related air pollution. Atmos Environ 46:313–319.

Zaharna M, Guilleminault C. 2010. Sleep, noise and health: a review. Noise Health 12:64–69.