Air pollution and general practitioner access and utilization: a population based study in Sarnia, ‘Chemical Valley,’ Ontario

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Abstract

Background: Health impacts of poor environmental quality have been identified in studies around the world and in Canada. While many of the studies have identified associations between air pollution and mortality or morbidity, few have focused on the role of health care as a potential moderator of impacts. This study assessed the determinants of health care access and utilization in the context of ambient air pollution in Sarnia, Ontario, Canada.

Methods: Residents of Sarnia participated in a Community Health Study administered by phone, while several ambient air pollutants including nitrogen dioxide (NO2), sulphur dioxide (SO2) and the volatile organic compounds benzene, toluene, ethylbenzene, m- and o-xylene (BTEX) were monitored across the city. Land Use Regression models were used to estimate individual exposures to the measured pollutants and logistic regression models were utilized to assess the relative influence of environmental, socioeconomic and health related covariates on general practitioner access and utilization outcomes.

Results: The results show that general practitioner use increased with levels of exposure to nitrogen dioxide (NO2- Odds Ratio [OR]: 1.16, p < 0.05) and sulphur dioxide (SO2- OR: 1.61, p < 0.05). Low household income was a stronger predictor of having no family doctor in areas exposed to high concentrations of NO2 and SO2. Respondents without regular care living in high pollution areas were also more likely to report travelling or waiting for care in excess of 20 minutes (OR: 3.28, p < 0.05) than their low exposure counterparts (OR: 1.11, p > 0.05).

Conclusions: This study provides evidence for inequitable health care access and utilization in Sarnia, with particular relevance to its situation as a sentinel high exposure environment. Levels of exposure to pollution appears to influence utilization of health care services, but poor access to primary health care services additionally burden certain groups in Sarnia, Ontario, Canada.

Background

Research based on recent conceptualizations of health that recognize socioeconomic and environmental determinants in Canada shows that significant health disparities continue to persist, despite a health care system based on the premise of universal and equitable access [1,2]. Regional health care expenditures in Ontario are also associated with toxic pollution output [3,4]. Such associations between the physical environment and health have highlighted the need to study the compounding impacts of environmental and socioeconomic stressors with a focus on context specific environmental health and health care issues.

Analyses of primary health care outcomes under the assumption of universal access have revealed that access and utilization depend on a host of individual, social and environmental factors. For example, physician use is determined by both individual and neighbourhood income in addition to educational attainment [5,6]. Dunlop et al. [7] found that lower socioeconomic status (SES) was associated with increased use of primary care and lower rates of specialist service utilization, whereas health needs as measured by perceived health status and health condition predicted utilization across the board. Finkelstein [8] found that health care expenditures were related to income, but this association disappeared after...
adjusting for health status. Curtis and MacMinn [9] indicated that inequities related to lower levels of income and education in Canada actually grew between 1978 and 2003, while health status remained strongly associated with health care utilization. However, SES inequities in utilization appeared attenuated after initial contact with the primary care system. Among studies that did not consider temporality, but rather focused on spatiality, efforts have included controlling for health service environments as different among Public Health Units [10], and both system- and individual-related barriers within neighbourhoods [11].

Birch et al. [12] argued that health care resource allocation should be based on the needs of a particular community rather than presenting patients, who in practicality form the basis for funding to a particular service provider. This stems from provider inability to determine the needs of those who do not present themselves as patients, which is a problem reflected in the geography of health care literature concerned with the spatial organization of health services. Geographic analysis of health care draws attention to the complexity of population needs in health service access and utilization [13]. The argument is that the relationships between individual, population and neighbourhood characteristics that determine need in a particular context (e.g., demographics, SES, mobility, ethnicity, pollution, etc.) need to be examined in order to understand health care issues. Equitable access to health care cannot be meaningfully discussed without considering needs [14].

Research on health care outcomes with explicit reference to the physical environment is most often conducted at the level of neighbourhood or intra-urban scales to capture adequate resolution of spatial variability [15,16]. Previous studies have established relationships between neighbourhood and population characteristics that amplify environmental health impacts, but to our knowledge no studies have examined the relationship between health care and air quality directly. However, existing literature indicates that people exposed to higher levels of air pollution are particularly vulnerable to the effects of inequitable access to primary health care [17]. Access to primary care helps prevent illness and death, and is also more highly associated with equitable distribution of health in populations than specialty care [18]. Gwynn and Thurston [19] suggest that disparate access to health care may increase susceptibility to the effects of air pollution. In a Canadian context, this is of particular interest since the Canada Health Act implicitly assumes that care is available to those who are in need, but individual providers have historically been trusted to allocate resources according to federal objectives [12]. Local health care providers are faced with resource restraints that can make it difficult to meet federal objectives, and in this study we seek to further understand how the distribution of air pollution is associated with consumption of health care in Sarnia, Chemical Valley, Ontario [20].

Study Context
The City of Sarnia has a population of 71,419 and covers approximately 800 km² [21]. The Sarnia area is called ‘Chemical Valley’ as it is home to more than 40 per cent of all chemical processing facilities in Canada. Sarnia is also located within the government designated St. Clair River Area Of Concern (AOC), which among 16 other areas in Canada was declared in further need of health investigations relating to impacts of environmental pollution [22]. The region is also subject to significant outputs of vehicular exhaust due to the Canada-US international crossing at Bluewater Bridge, and transnational air and water pollution from Ohio, Illinois and Michigan. The Ontario Medical Association [4] estimated that Sarnia and the surrounding Lambton County suffered 100 premature deaths, 270 hospital admissions, 920 emergency visits and 471 700 minor illness days due to air pollution in 2005 alone. Fung and colleagues [20] showed that rates of hospitalization were significantly higher in Sarnia than Windsor and London, two nearby cities in Southwestern Ontario. Furthermore, a study focused on the Aamjiwnaang First Nation’s reserve, which is surrounded by Chemical Valley (Figure 1), reported a sex ratio of 2:1 in favour of females and attributed this anomaly to accumulative effects of pollution [23,24]. This background information calls for the need to examine the relationship between air pollution and health care utilization at the community level. The findings will provide guidance to local level health promotion and preventive care deficiencies with a focus on areas with poor air quality.

Methods
We use a combination of spatial, environmental, and survey data to generate and test the hypothesis that relationships between air pollution and determinants of health, access and utilization of general practitioners (GP) do exist in Sarnia, Ontario. Residents of Sarnia took part in a community health study in 2005. A stratified random sampling procedure was used to select respondents from each census tract. The survey was conducted by Canadian Viewpoint Ltd, a survey company in Toronto, Ontario (http://www.canview.com) using a computer assisted telephone interview system and introduced as a general health survey. The sample represented approximately 1% of the Sarnia population, yielding a total of 804 respondents with a response rate of 62%. This sample size was determined to be sufficiently large for the number of variables included in the
analyses [25]. The study was approved by the Social Sciences Research Ethics Board at The University of Western Ontario.

**Outcome Variables**

Access to health care was measured with the question “Do you have a regular family medical doctor or health care provider?” We recognize that health care consumers have over the last decade become more dependent on primary care sources other than family doctors (e.g., nurse practitioners, walk-in clinics). Hence this conceptualization of access may be a limited measure of the potential for health care use. Nonetheless, access to a regular medical doctor is supposed to be equitable and represents a particular aspect of quality of care, because continuity of regular care can strongly influence health care satisfaction and use of health care services [26]. Furthermore, based on the important distinction between this definition of access versus the act of using or receiving health care [27], we measure utilization of primary health care, or general practitioner (GP) use, by a positive response to the question “in the past two weeks, have you seen or talked on the telephone with your family doctor about your physical, emotional or mental health?” Similar questions have been used in previous health care utilization studies [16], and Roberts et al. [28] showed that this self-reported measure is reasonably accurate. Therefore we used these questions to evaluate short-term effects of air pollution on health care utilization and equitable access to services, and furthermore identify the inherent mediating factors that contribute to both.

**Independent Variables**

**Exposure Assessment**

During administration of the survey several ambient air pollutants including nitrogen dioxide ($\text{NO}_2$), sulphur dioxide ($\text{SO}_2$) and the volatile organic compounds
benzene, toluene, ethylbenzene, mp- and o-xylene (BTEX) were monitored at 39 locations across the city of Sarnia for 2 weeks in October 2005 [29,30]. Land use regression (LUR) was utilized to model the ambient NO2, SO2 and total BTEX concentrations because of its potential to provide spatial pollution estimates in small-areas without the data requirements and related expenses incurred in other exposure modelling techniques, such as dispersion [31] and micro-environmental monitoring [32]. LUR modelling uses nearby traffic, land use and population variables to explain the spatial variability of air quality [33]. For a comprehensive review of LUR models for exposure assessment in epidemiologic studies see Hoek et al. [34]. We controlled for personal exposure by asking the respondent what number of outdoor (gases, dusts, fumes, and pesticides) or indoor (pets, carpets, rugs and fireplaces) irritants to which they were regularly exposed. Table 1 provides a description of all variables that were included in the analysis.

Community Context
Respondents were asked to assess their degree of annoyance due to air pollution odours on an 11-point scale as similar measures have previously been used to study annoyance [35]. Respondents were also asked if they believe the odours from the chemical plants were harmful to their health and whether they were aware of Sarnia being classified as an AOC. These were dichotomous variables with those who answered “no” used as the reference categories. A categorical variable constructed from travel and wait times for health care provided a measure of availability and those who responded with “20 minutes or less” for either waiting or traveling were used as the reference category [16]. Respondents were also asked to rate their level of satisfaction with Sarnia as a community. The variable was dichotomized with the first two responses of the following question defined as the outcome of interest - “In general, how satisfied are you with your community as a place to live? Would you say you are Very Satisfied, Somewhat Satisfied, Not Too Satisfied, or Not At All Satisfied?”

Health and behaviour
Chronic health was included as a continuous variable of the number of conditions diagnosed by a physician and included skin conditions, respiratory disease, hypertension and cancer. General symptoms were defined as those likely to be caused by stress-mediated mechanisms of pollution and included chest pain, headaches, dizzy spells, sleep problems, stomach aches, diarrhea and loss of appetite. Cardinal symptoms were defined as those likely to be the result of irritant properties of air pollution and included coughs, wheezing/breathing problems, nausea, sinus congestion, colds, skin rashes, eye, nose or throat irritations, earaches and nosebleeds [36]. Control symptoms (backpain, joint pain and easy bruising) not likely related to air pollution were also included. Mental health was measured using the General Health Questionnaire (GHQ - 20 items) [37]. Those who provided positive responses to at least four of the questions were categorised as emotionally distressed [38]. Respondents’ self-rated health compared to others their age was classified as poor or fair versus good, very good or excellent. Coping was classified the same way as health status, but by response to the question “how would you rate your ability to handle day-to-day demands in your life?” Measures of health related behaviour including smoking, alcohol consumption, exercise and medical check-ups were also included.

Socioeconomic and Demographic Characteristics
Age was included as a categorical variable with the intervals 18-24, 25-44, 45-64, and 65 or older, while males were used as the reference category for the influence of sex. Income below the Statistics Canada Low Income Cut-Off [39] at $22,139 before tax for the median number of household members in our sample and population size in Sarnia, having no children under age 18, marital status, completion of high school and employment were used as reference categories for variables measuring family characteristics and SES. Housing condition was coded as “satisfactory” vs. “in need of repair”, while housing tenure was coded as “owning” vs. “renting”. These housing variables were included in the analysis to control for their potential influence on health and as an additional SES measure (e.g., [40]).

Analysis
Binary logistic regression models for the dependant variables were built in which each consecutive, conceptually grouped, block of variables were entered into a stepwise regression algorithm (SES, health and behaviour, community context, and pollution health effects) following forced entry of a priori variables. The a priori variables were chosen because of their influence as demonstrated in previous studies and included age, sex, family doctor, chronic disease and mental health in the GP use model [41] and income, age, sex, unmet health care needs in past 12 months and self rated health [11,42] in the GP access model. The variables that made a significant contribution to the model at the entry of each block were retained and included in the next block. As suggested by Hosmer and Lemeshow a less stringent significance level of 0.15 for inclusion was utilized to ensure variables with coefficients different from zero were included [43].

Based on our study objectives, we examined if burdens of air pollution were borne by residents of Sarnia who were also at a disadvantage with respect to health care access. Therefore, a priori and other significant
| Variable                                      | Type          | Coding                                      |
|----------------------------------------------|---------------|---------------------------------------------|
| **Demographic**                              |               |                                             |
| Age of respondent                            | Categorical   | 18-24* vs 25-44, 45-64, 65+                |
| Gender                                       | Categorical   | Male* vs Female                             |
| **Exposure**                                 |               |                                             |
| Nitrogen dioxide (NO₂)                       | Continuous    |                                             |
| Sulphur dioxide (SO₂)                        | Continuous    |                                             |
| Benzene, toluene, ethylbenzene and xylenes (BTEX) | Continuous |                                             |
| Gas, dust, fumes and pesticides (outdoor exposure) | Continuous |                                             |
| Pets, carpets, rugs and fireplace (indoor exposure) | Continuous |                                             |
| **Socioeconomic**                            |               |                                             |
| Marital status                               | Categorical   | Partner* vs No partner                      |
| High school graduate (education)             | Categorical   | Yes* vs No                                  |
| Below Low Income Cut Off                     | Categorical   | No* vs yes                                  |
| Employment status                            | Categorical   | Employed*, Other, Unemployed                |
| Children under 18 (children)                 | Categorical   | No* vs yes                                  |
| Housing tenure                               | Categorical   | Owned* vs rented                            |
| Housing condition                            | Categorical   | Satisfactory* vs needs repair               |
| **Health Behaviour**                         |               |                                             |
| Frequency of monthly alcohol use             | Continuous    |                                             |
| Regular smoker                               | Categorical   | No* vs Yes                                  |
| Hours of weekly exercise                     | Continuous    |                                             |
| Voluntary Medical Check-up                   | Categorical   | No* vs Yes                                  |
| **Community & Environmental Perception**     |               |                                             |
| Odour annoyance                              | Continuous    | 0-10                                        |
| Aware that Sarnia is considered in an Area of Concern (awareness) | Categorical | No* vs Yes                                  |
| Believe odours/pollution cause health problems (pollution health) | Categorical | No* vs Yes                                  |
| Community satisfaction                       | Categorical   | Dissatisfied* vs Satisfied                  |
| **Health Care**                              |               |                                             |
| Health care provider travel and wait times (Care availability) | Categorical | ≤ 20 min vs > 20 min                        |
| **Health Status**                            |               |                                             |
| Physician diagnosed chronic health problems (chronic conditions) | Continuous |                                             |
| Health problems from pollution-induced stress (stress symptoms) | Continuous |                                             |
| Cardinal health problems of pollution (cardinal symptoms) | Continuous |                                             |
| Back- or joint-pain and easy bruising (control symptoms) | Categorical | None* vs any                                |
| Above cut-point 4 positive responses on General Health Questionnaire | Categorical | No* vs Yes                                  |
| Self-rated health compared to other people same age | Categorical | Good or better* vs Fair/poor               |
| Ability to cope with daily problems (coping) | Categorical   | Good or better* vs Fair/poor               |
| **Outcomes**                                 |               |                                             |
| **Health Care Access**                       |               |                                             |
| Regular health care provider or family physician (GP Access) | Categorical | Yes* vs No                                  |
| **Health Care Utilization**                  |               |                                             |
| General practitioner use in the past 2 weeks (GP Use) | Categorical | No* vs Yes                                  |

* Reference category.
independent variables remaining in the final models for GP use and access were forced into regression models that assessed the impacts of the monitored pollutants. Power transformations were applied to several of the continuous variables for the regression analysis as their distributions were skewed (Figure 2). To evaluate the combined impact of pollutants each respondent was subsequently ranked based on estimated exposure to NO2 and SO2, the results of which were in turn summed and ranked to assign an overall exposure rank to individual respondents (1-804) [44]. We split the sample into tertiles of low, medium and high rank in order to compare the determinants of health care access and utilization in spatially representable sub-samples (Figure 1). The number of sub-samples was limited to three because of sample size restrictions and the number of variables included in the analysis. All analyses were carried out with SPSS 17, PASS 11 and ArcGIS 10.0.

Results
Sample Characteristics
Table 2 shows the observed frequencies of predictors within the dependant variable classes and the entire sample. Comparing our sample parameters to census data [21] showed that females were slightly over-represented and with respect to age the sample represented an older population than in Sarnia at large. The sample characteristics suggest that the sub-population of Sarnia without a family physician or a regular care provider is disproportionately composed of younger males. This is consistent with reports based on the 2005 Canadian Communities Health Survey data, which found that older people and females used health care services more often than their young and male counter parts [41]. Perceived health status as measured in Statistics Canada’s community health profiles [45] was comparable with mental health measured by the GHQ was the only a priori variable that did not contribute to the initial model, while sex (females), age, chronic disease, and GP access significantly increased the odds of GP consultations. These variables remained significant predictors of GP use throughout the analysis, except that gender was insignificant when odour annoyance and community satisfaction were introduced into the model. This could be due to females reporting significantly higher levels of odour annoyance (Anova $F = 25.05, df = 1, p < 0.001$). The effect of odour annoyance disappeared when pollution related stress symptoms were entered into the model and females also reported significantly higher levels of stress.

The final models show that pollutants have different impacts on GP use. Table 4 shows the parsimonious final model for GP use compared to 4 other models that only differ by having SO2, NO2, BTEX or odour annoyance individually entered. Higher levels of exposure to NO2 and SO2 significantly increased the likelihood of seeing a doctor by 16-60% during the 2 weeks that monitoring took place. The results produced from this analysis are not very strong potentially due to the relatively low power of the models that included the different pollutants. Only the model that included odour annoyance had higher power than the final stepwise model. The influence of GP access was similar in the control and NO2 models, but diminished with SO2 and BTEX. Comparing the odds of other variables between the control and pollutant models revealed that the contribution of age and stress also changed markedly. Specifically, there was an increased influence of stress with SO2 that attenuated the impact of age. The Wald statistics confirmed that stress proxy made a higher contribution to the model controlling for SO2 than NO2. Interestingly, we also observed that the impact of smoking on GP use decreased with NO2 and BTEX but became insignificant when controlling for SO2.

Respondents who reported higher satisfaction with their community were approximately 50% more likely to have seen or spoken with their family doctor in the control and all pollution models. We looked to the models measuring access for a possible chain relationship, but entry of the community satisfaction variable into the access model resulted in over-fitting. However, bivariate

Health Care Utilization
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Figure 2 Box-plot diagrams of NO2, SO2, and BTEX values for respondents based on Land Use Regression Models within categories of general practitioner (GP) utilization and regular GP Access. Solid lines across panels represent pollutant average and inside boxes the median. (○): 1.5-3 times Interquartile range; (★) 3 or more times interquartile range.
Analysis revealed that there was a significant and directional level of association between access and the ordinal measure of community satisfaction as the dependent variable (Somers’ D = 0.31, p ≤ 0.001). This suggests that community satisfaction is influenced by access to regular care that subsequently affects utilization.

Access to Primary Health Care

All a priori (sex, age, income, unmet health care needs and self-rated health) variables were significant and together provided a model of similar strength to the final GP use models (Table 5). When SES and health related measures were entered into the stepwise procedure the influence of sex and health care needs changed significantly. Males were almost twice as likely as females to have no family physician after SES variables were included, but this effect disappeared with the entry of health and behavioural measures. This was likely due to higher rates of back-pain, joint-pain and easy bruising among women in our sample. Older groups of the sample were more likely to have GP access and this effect remained significant throughout the analysis. Removal of the variable measuring medical check-ups rendered age an insignificant predictor within the entire sample, but in the models that compared high, medium and low exposure sub-samples, only respondents aged 45-64 in the high rank sample had significantly better primary care access.

Analysing the exposure ranked sub-samples identified notable relationships between air pollution and health care. The low exposure sub-sample had the same power as the full sample followed by the high and medium rank samples. High exposure respondents faced stronger barriers to regular primary care when compared to those in lower pollution areas based on several of our independent variables. For instance, we observed that the LICO only predicted access significantly within the medium and high exposure zones and not the low exposure area (Figure 1). Also, respondents living in rental housing within the high exposure area were more likely to lack GP access (OR: 6.02) than low exposure respondents (OR: 4.29) and medium exposure respondents (OR: 2.24). Further analysis of these findings by age distribution showed that there was a significant interaction between housing tenure and exposure rank groups (F = 6.96, df = 2, p < 0.01), possibly owed to the clustering of retirement homes and post-secondary schools. Significant influences of location and wait-times for health care services were restricted to the medium and high exposure sub-samples. Although respondents who spent more than 20 minutes travelling and/or waiting for care were twice as likely to lack regular care overall, this OR was 4.5 in the medium rank, 3.2 in the high rank and non-significant in the low exposure area. The control symptoms did not significantly contribute to GP access in any of the exposure zone sub-samples.

Residual analysis identified a small number of outlying cases with high leverage or influence (DfBeta) values, but the variable coefficients or their significance did not notably differ when these outliers were removed from...
the analysis. We present McFadden $R^2$, or rho-square values, which between 0.2-0.4 represent a very good fit of the model. The pseudo $R^2$ values and model scores indicated that the GP access models were in general stronger than for GP use. This is because it is much more difficult to capture the innumerable reasons for visiting a GP [47] versus the population characteristics associated with health care access. The Hosmer-Lemeshow chi-square statistics indicated that the goodness of fit in the final models were satisfactory.

Discussion
The study found that multiple dimensions of health along with environmental, socioeconomic, demographic, and health service spatiality predicted GP access and utilization in Sarnia. The importance of differentiating access and utilization, and furthermore taking a closer look at how they interact is exemplified in a high exposure environment. Within the context of compromised environmental quality, our findings demonstrate that neighbourhoods, individual and environmental characteristics can interact to predict access and utilization of health care. Besides the apparent impact of access to regular care on community satisfaction, our analysis revealed that the \textit{a priori}, SES, health care needs and other health related variables were significantly associated with GP access. These results confirm findings from previous studies in Canada and abroad [48]. For instance, the findings here support the OMA’s report [4] on the increased cost of health care due to air pollution in the Sarnia region. The results demonstrate that there are significant and specific impacts of different pollutants and their spatial distributions on primary care access and utilization.

A previous study in Sarnia found that high annoyance scores were significantly related to both NO$_2$ and SO$_2$ levels [35]. Other studies have found associations between NO$_2$, cardiac autonomic dysfunction and increased blood pressure, which are also symptoms of stress and can lead to cardiovascular disease [49,50]. NO$_2$ can also increase the risk of respiratory tract infections through interaction with the immune system and SO$_2$ contributes to respiratory problems for both healthy subjects and those with pulmonary disease [51]. Increased likelihood of GP use from smoking cigarettes was attenuated by NO$_2$ and BTEX, and the effect became insignificant with SO$_2$ in the model. Therefore, we suggest that NO$_2$, BTEX and SO$_2$ in Chemical valley increase health care utilization because of their effects on respiratory and cardiovascular diseases prominently.

| Table 3 Stepwise Logistic Regression Model for GP Use | A priori | SES | Health & Behaviour | Community Context | Pollution Health Effects (CI) |
|------------------------------------------------------|---------|-----|-------------------|-------------------|-------------------------------|
| Sex (Female)                                         | 1.514*  | 1.532* | 1.530*           | 1.426             | 1.295 (0.893-1.878)        |
| Age (18-24)                                          |         |       |                   |                   |                               |
| 25-44                                                | 2.953*  | 3.790* | 3.560*           | 3.313*           | 3.690* (1.201-11.342)      |
| 45-64                                                | 2.398   | 3.246* | 2.932            | 2.675             | 3.162* (1.204-9.763)       |
| 65+                                                  | 2.339   | 3.100* | 3.147*           | 3.047             | 4.306* (1.330-13.938)      |
| Chronic Conditions                                   | 1.193***| 1.194***| 1.210***         | 1.207***         | 1.131* (1.020-1.256)      |
| GHQ 4+                                               | 1.225   | 1.194  | 1.199            | 1.177             | 1.046 (0.696-1.574)        |
| GP Access                                            | 2.705*  | 3.235* | 3.388*           | 2.978*           | 3.091* (1.141-8.374)      |
| Housing Tenure                                       |         |       | 1.857***         | 1.819**           | 1.844** (1.279-2.723)     |
| Smoking                                              |         |       |                   |                   |                               |
| Medical Check-up                                     | 1.611*  | 1.546* | 1.622*           | 1.060-2.484       |
| Community Satisfaction                               |         |       |                   |                   |                               |
| Odour Annoyance                                      | 1.068*  |       | 1.052            | (0.995-1.113)     |
| Stress Symptoms                                      |         |       |                   |                   |                               |

Diagnostics

| Model Likelihood Ratio | $\chi^2 = 37.48$, df = 7, p < 0.001 | $\chi^2 = 48.67$, df = 8, p < 0.001 | $\chi^2 = 58.61$, df = 10, p < 0.001 | $\chi^2 = 66.69$, df = 12, p < 0.001 | $\chi^2 = 78.13$, df = 13, p < 0.001 |
|-----------------------|-------------------------------------|-------------------------------------|-------------------------------------|-------------------------------------|-------------------------------------|
| McFadden $R^2$        | 0.05                                | 0.06                                | 0.07                                | 0.11                                | 0.11                                |
| Hosmer-Lemeshow Test  | $\chi^2 = 12.05$, df = 8, p = 0.15  | $\chi^2 = 4.35$, df = 8, p = 0.82  | $\chi^2 = 6.80$, df = 8, p = 0.56  | $\chi^2 = 16.00$, df = 8, p = 0.04  | $\chi^2 = 12.55$, df = 8, p = 0.128 |

Significance level: * < 0.05, ** < 0.01 *** < 0.001.
Table 4 Logistic Regression Models of GP Use with NO₂, SO₂, BTEX and Odour Annoyance

|                      | No Pollution | NO₂   | SO₂   | BTEX  | Odour Annoyance |
|----------------------|--------------|-------|-------|-------|-----------------|
|                      |              | 1.155*| 1.611*| 1.407 | 1.060*          |
| **Age (18-24)**      |              |       |       |       |                 |
| 25-44                | 3.765*       | 3.625*| 2.821*| 2.969*| 3.707*          |
| 45-64                | 3.234*       | 3.195*| 2.459*| 2.622 | 3.115*          |
| 65+                  | 4.242*       | 4.206*| 3.218*| 3.485*| 4.390*          |
| **Chronic Conditions**| 1.139*       | 1.128*| 1.134*| 1.134*| 1.133*          |
| **GP access**        | 4.318**      | 4.516**| 3.371*| 3.432*| 3.150*          |
| Housing Tenure       | 1.932*       | 1.703**| 1.778**| 1.858***| 1.879***        |
| Smoking              | 1.613*       | 1.564*| 1.471 | 1.529*| 1.581*          |
| Medical Check-up     | 1.679*       | 1.661*| 1.603*| 1.640*| 1.633*          |
| Community Satisfaction| 1.536*       | 1.535*| 1.604*| 1.599*| 1.639*          |
| **Stress Symptoms**  | 8.523***     | 8.826***| 9.126***| 8.946***| 7.402***        |

Significance level: * < 0.05, ** < 0.01 *** < 0.001.

Model Likelihood Ratio \( \chi^2 = 74.81, \text{df} = 10, p < 0.001 \)

McFadden R² 0.09

Hosmer-Lemeshow Test \( \chi^2 = 11.79, \text{df} = 8, p = 0.16 \)

Correctly Classified (\%) 76.7

Table 5 Stepwise Logistic Regression Model for GP Access (no regular care)

|                      | A priori | SES | Health & Behaviour (CI) |
|----------------------|----------|-----|-------------------------|
| **Sex (female)**     | 1.993*   | 1.865*| 1.596 (0.869-2.932)    |
| **Age (18-24)**      | **       | *   | *                       |
| 25-44                | 904      | 1.199| 1.388 (0.532-3.681)    |
| 45-64                | 387*     | 540 | 648 (0.233-1.802)      |
| 65+                  | 280*     | 313*| 367 (0.113-1.194)      |
| **Low Income**       | 4.043*** | 2.878**| 3.001** (1.411-6.384)  |
| **Unmet Needs**      | 3.580*** | 3.904***| 4.313*** (2.116-8.790)|
| **Self-Rated Health**| 2.722*   | 3.426*| 3.308* (1.270-8.616)   |
| **Education**        | 2.619*   | 2.575*| 1.097-8.048            |
| **Housing Tenure**   | 2.563**  |      | 2.458** (1.275-4.739)  |
| **Control Symptoms** | 1.891*   | 1.04-3.561 |
| **Medical Check-ups**| 1.719 | 0.929-3.182 |
| **Care Availability**| 2.268** | 1.240-4.149 |

Significance level: * < 0.05, ** < 0.01 *** < 0.001.

Model Likelihood Ratio \( \chi^2 = 46.38, \text{df} = 7, p < 0.001 \)

McFadden R² 0.12

Hosmer-Lemeshow Test \( \chi^2 = 6.25, \text{df} = 7, p = 0.51 \)

Significance level: * < 0.05, ** < 0.01 *** < 0.001.
featured in the area [20]. Thompson et al. [52] reported that daily fluctuations in benzene concentrations predicted acute asthma emergency admissions of children in Belfast, Northern Ireland, but our analysis showed no direct significant relationship between BTEX and our outcomes. This may be because the predicted distribution of these compounds when taken together does not reveal their individual impacts since their modes of dispersion and the way people respond to them differ. Most health effects associated with VOCs are observed over longer periods of time.

Our results suggest that low SES and stress enhanced by odour annoyance may be compounding health care demands, possibly due to increased susceptibility to adverse pollution impacts (e.g., compromised immune system) [53]. This ‘double burden of deprivation’ has been identified in studies conducted in Worcester, U.S. A., and Montreal, Canada [15,54]. We observed that the low income only predicted access significantly within the medium and high exposure zones and not the low exposure area. This finding signifies a worrisome interaction between burdens of social and environmental stress, and access to primary health care services. We also found that high exposure ranked respondents were more likely than low ranked respondents to lack GP access if renting, which provides further support for low SES being a barrier to health care.

The role of odour annoyance as a stressor in Sarnia is consistent with work by Shusterman et al. [55] who found that odour mediated mechanisms and annoyance contribute to how people judge and cope with air quality, and furthermore provide important diagnostic information in appraising the potential threats to health and well-being. Research on beliefs regarding toxicity of environmental pollution suggests that “if environments smell bad, they’re probably damaging to health” [56] or at the very least, they may reinforce annoyance. Consistent with earlier studies [57], we found females were more likely to report high odour annoyance than their male counterparts. This could be because of gender differences in cognitive and affective processes [57].

Negative perceptions about the environmental health of one’s neighbourhood can influence health outcomes along with predisposing determinants of health [58,59]. Consequently, perceived personal susceptibility and severity of health threats are modified by psychosocial factors, which thereafter can influence compliance with medical recommendations and perceived benefits of preventive action [60]. Benefits of access to regular care are not only preventive, but also associated with the responsibility placed on health care providers to inform and educate their patients about hazards of air pollution [61]. It is therefore particularly concerning to find that residents in Sarnia who live in high exposure areas spend more time travelling and waiting for GP consultations. This may be due to primary care providers in Sarnia locating their practices in less polluted areas, thus leaving those in highly polluted areas to travel long distances to seek care. Further research is required to determine the apparent challenges of delivering primary health care services to these areas and their populations.

Although levels of estimated exposure for respondents in the current study were within provincial guidelines, our monitoring methods were not able to capture the impact of individual release events of airborne toxins or “bad air days”; they are not uncommon and are occasionally accompanied by warning sirens and emergency response guides that demand individual coping mechanisms [62]. We can assume that residents at risk from chronic exposure are the same residents at most risk from these events. Residential areas within the high rank zone of Figure 1 provides target areas for improving health care delivery as population characteristics in this zone predicted inequitable access and coincide with high NO₂ and SO₂ concentrations that increased the likelihood of utilizing services.

There are a few limitations to this analysis that are worth noting. We note that our conceptualization of access related to a source of regular care at the time of the study and did not account for barriers to acquiring a family doctor or the fact that some people do not attempt to find regular care. Our measure of access did not include alternative sources of care such as walk-in clinics, Telehealth Ontario (free telephone consultation with a registered nurse) and nurse practitioners, and these providers arguably represent important points of access for certain deprived populations identified in this study. A study that looked at access to family physicians in Southwestern Ontario found that of the 9.1% of the population that did not have regular care, 55% used walk-in clinics and 13% used emergency rooms as their source of care [63]. The study was also limited by lacking longitudinal measures of health and SES as the study used a cross-sectional design. O’Neill et al. [64] suggest studies on air pollution that include SES measures consider how they change through the course of life. They also propose that exposure assessment include the effect of daily movement, which our study design did not permit. Furthermore, we found that our model predicting health care utilization was relatively weak, but there were nonetheless significant associations with air pollution that have potential implications for policy.

Conclusions

The aim of the study was to examine the relationship between spatially sensitive measures of air pollution and general practitioner access and utilization in a high exposure environment. Our results provide further
support to the pursuit of better access to regular primary care in communities faced with environmental challenges. The Ontario Medical Association projects that increased numbers of premature deaths, hospital admissions and emergency visits due to air pollution in Lambton County, which includes Sarnia and smaller communities in the surrounding area, will increase to $6 million in lost productivity and $8 million in additional health care costs by 2026. We conclude that some of these costs can be avoided by ensuring equitable access to primary care for residents most severely affected by air pollution.

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Authors’ contributions

THO and IL conceived the study and were involved in interpretation of the results and preparation of the manuscript. DA and KG were involved in the analysis and all authors have read and approved the final manuscript.

Competing interests

The authors declare that they have no competing interests.

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Abbreviations

AOC: Area of Concern; BTEX: Benzene, toluene, ethylbenzene, xylene; CI: Confidence Interval; GP: General Practitioner; LUR: Land Use Regression; NO2: Nitrogen dioxide; OMHA: Ontario Medical Association; OR: Odds Ratio; SES: Socioeconomic Status; SO2: Sulphur dioxide; VOC(s): Volatile Organic Compound(s).
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