PM$_{2.5}$ decadal data in cold vs. mild climate airports: COVID-19 era and a call for sustainable air quality policy

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Abstract

Airports are identified hotspots for air pollution, notably for fine particles (PM$_{2.5}$) that are pivotal in aerosol-cloud interaction processes of climate change and human health. We herein studied the field observation and statistical analysis of 10-year data of PM$_{2.5}$ and selected emitted co-pollutants (CO, NO$_x$, and O$_3$), in the vicinity of three major Canadian airports, with moderate to cold climates. The decadal data analysis indicated that in colder climate airports, pollutants like PM$_{2.5}$ and CO accumulate disproportionally to their emissions in fall and winter, in comparison to airports in milder climates. Decadal daily averages and standard errors of PM$_{2.5}$ concentrations were as follows: Vancouver, 5.31 ± 0.017; Toronto, 6.71 ± 0.199; and Montreal, 7.52 ± 0.023 μg/m$^3$. The smallest and the coldest airport with the least flights/passengers had the highest PM$_{2.5}$ concentration. QQQ-ICP-MS/MS and HR-S/TEM analysis of aerosols near Montreal Airport indicated a wide range of emerging contaminants (Cd, Mo, Co, As, Ni, Cr, and Pb) ranging from 0.90 to 622 μg/L, which were also observed in the atmosphere. During the lockdown, a pronounced decrease in the concentrations of PM$_{2.5}$ and submicron particles, including nanoparticles, in residential areas close to airports was observed, conforming with the recommended workplace health thresholds (~2 × 10$^4$ cm$^{-3}$), while before the lockdown, condensable particles were up to ~1 × 10$^5$ cm$^{-3}$. Targeted reduction of PM$_{2.5}$ emission is recommended for cold climate regions.

Keywords

Airport pollution · Air quality · Particles in snow · Aerosols · Emerging contaminants · COVID-19

Introduction

The World Health Organization (WHO) considers that air pollution kills approximately 7 million people every year worldwide (WHO 2021). Air pollution is composed of gases and suspended particles, or aerosols with diameters of a few nm to several microns. In this paper, we refer to aerosols with a diameter less than 1 micron as submicron, which includes nanoparticles. Aerosols are detrimental to human health and the Earth’s climate (Organization 2010; Stocker 2014). Of all air pollutants, smaller aerosols (<100 nm aka nanoparticles) are of major concern for human health. The WHO considers airborne nanoparticles a major cause of premature human mortality in children (Nazarenko et al. 2021; Organization 2014).

The effects of particles on human health depend on their different physical and chemical properties and exposure time. For instance, the deposition of particles in the respiratory system depends on the size of the inhaled particles (Kreyling et al. 2006). Particles of smaller sizes can travel further down the respiratory system (Kreyling et al. 2006). Nanoparticles are considered to have more toxicity on human health as they can be deposited in the alveolar region of the pulmonary system (Yang et al. 2008). Airborne nanoparticles are also the most numerous aerosols (~80%) (Morawska et al. 2008).

Airport-related emissions are the upmost contributors to air pollution in areas near airports (Chauhan and Singh 2020, Mazaheri et al. 2011, Singh and Chauhan 2020), where a vast increase in particulate matter (PM) is observed (Unal et al. 2005). It has been reported that ~97% of particles are due to aviation activities, including fine particles or particulate matter with diameters smaller than 2.5 microns...
Exposure to effects in people working at airports (Bendtsen et al. 2021). Airplane contrails contain soot and other particles emitted from aviation activities, on which water vapor condenses (Kärcher 2018). In airports, a significant source of PM$_{2.5}$, including soot and other carbonaceous aerosols, is fuel burning by jet engines (Stacey 2018). Notwithstanding that airplane fuels are different motor vehicles, mostly using kerosene instead of gasoline fuel, and thus different amounts and types of pollutants (Midcontinent 2021).

Aircraft emissions of PM$_{2.5}$ include soot (Timko et al. 2010), metals, and other organic and inorganic emerging contaminants (Masiol and Harrison 2014, Rahim et al. 2019). Aside from particles, engines exhaust from aircraft produces several toxic gases such as NO$_x$, CO, SO$_2$, and other greenhouse gases (Bhattacharjee et al. 2015; Harrison et al. 2015). Airborne particles related to commercial aviation have been shown to be emitted predominantly during aircraft takeoff, landing, or idling activities, thus affecting airport regions (Camero 2019; Mazaheri et al. 2011; Vennam et al. 2017). The aircraft idling practice is not only harmful to human health, but also it is expensive, it can damage the engine, and it is ineffective to warm up the engine (Bern 2019). It has been observed that exposure to particles generated by idling causes inflammation of the bronchial epithelial cells (Jonsdottir et al. 2019).

Jet engine emissions have been associated with health effects in people working at airports (Bendtsen et al. 2021). Exposure to PM$_{2.5}$ from aircraft emissions has shown short-term (Lammers et al. 2020) and long-term (Wing et al. 2020) health effects even on healthy individuals. Annual fossil fuel energy usage by the air traffic industry is estimated to be ~3% (Barrett et al. 2010). Black carbon particle emission, which is essential in climate change, is estimated to be approximately (10.9 ± 2.1) × 10$^{15}$ annually, with an average emission index of (6.06 ± 1.18) × 10$^{14}$ per kg of burned fuel for aviation activities (Zhang et al. 2019), whereas they were estimated for ~2.5% of global carbon dioxide (CO$_2$) emissions. It is notable that the impact of various other pollutant atmospheric distributions such as volatile sulfate and nitrate aerosols perturbed by aviation emissions is uncertain (Brasseur et al. 2016).

A major concern regarding air pollution from airports is the rapid increase of commercial flights every year due to the increased accessibility prior to the COVID-19 pandemic (Bhattacharjee et al. 2015, Chauhan and Singh 2020, Chauhan and Singh 2021). The aviation industry contributes 4% to global warming and 2.4% to CO$_2$ global emissions (Klöwer et al. 2021). Since 2004, there has been a global increase in the number of flights of about 69.3% (from 23.8 million in 2004 to 40.3 million in 2020 before COVID-19) with an increase of 3.38% ± 0.636 every year (Statista 2021a). Canada has followed the same trend, as shown in Fig. S5, illustrating the increase in the total number of passengers in Canada’s three major airports from 2008 to 2019. Toronto’s airport is the largest in Canada, followed by Vancouver (Chan 2020) and Montreal (Montreal 2020), with more than 417,228 flights in 2019 (Pearson 2020). In terms of the number of passengers, there has been an increase of 136.9% globally, from 1994 million in 2004 to 4723 million in 2020 (Statista 2021b). In Toronto’s airport, the number of passengers in 2014 was ~40 million and is expected to increase to ~80 million by 2035 (Pearson 2016). Nevertheless, there has been little to no regulation on air pollution emission by aircraft (Harrison et al. 2015). In 2020, however, due to the COVID-19 pandemic, global air traffic has seen a drastic decrease (Statista 2021c), resulting in decreased aircraft-related emissions (Berman and Ebisu 2020). According to a previous study, if the trends prior to COVID-19 resume, it is expected that the global temperature will increase by 0.1 °C by 2015 by the effects of the aviation industry alone (Klöwer et al. 2021).

This research had two goals; (I) evaluate whether 10-year statistical observations of PM$_{2.5}$ in the three largest Canadian airports in cold and mild climates are distinct, and how meteorological factors such as the cold temperature and snowfall in airport areas affect the composition of pollutants in snow, nearby airports; (II) determine whether a pronounced decrease in the concentrations of particles in residential areas close to airports was observed and whether the values are above or below the recommended health thresholds, during the COVID-19 lockdown.

**Experimental section**

**Particle size measurement**

We used a condensation particle counter (CPC) model 3007 (TSI Inc.) for field measurements. The CPC does not provide the particle size aggregated information, yet it is a condensation particle counter, providing the number density of particles from 0.01 up to 1.0 µm (TSI 2021).

**Sampling location for CPC**

The CPC instrument was operated periodically on operation days only from 0.01 to 1.0 µm. Note that equipment can be operated using batteries for two continuous hours. However, the battery can be switched, and the wick can be replenished every 2 h in a public park (Westwood Arena, 750 Avenue Thorncrest, Dorval, QC H9P 2R5) with a community sports center near the airport of Montreal shown in Fig. S1 in the supplementary information, a map of this location has been given.
Snow sample collection

Snow samples were collected near the airport of Montreal in a residential area at 750 Avenue Thorncrest, Dorval, QC H9P 2R5. A travel blank of Milli-Q water was taken for every sampling event. Samples were collected in pre-cleaned gas-tight glass amber jars using sterile equipment and clean suits. Snow samples were collected within 24 h of snowfall from the upper 3 cm of the snowpack. After collection, samples were transported in a cooler packed with snow (for isolation) to the laboratories at McGill University and stored immediately in a freezer at −35 ± 2 °C. Samples were thawed at room temperature just before analysis in clean conditions.

High-resolution scanning transmission electron microscopy with energy-dispersive X-ray spectroscopy (HR-STEM-EDS) and scanning electron microscopy (SEM)

We used HR-STEM-EDS to study the morphology, size, and elemental analysis of particles in snow collected near the airport of Montreal. For analysis, 5 mL of melted snow were deposited on a glow discharged carbon film–coated copper electron microscopy grid. The melted snow was left on the grid for 1 min and then blotted from the grid with the edge of a Kimwipe. Imaging was performed using a FEI Quanta 450 Environmental Scanning Electron Microscope (FE-STEM) and a FEI Tecnai G2F20 200 kV Cryo-STEM (FEI Electron Optics) equipped with a tungsten filament, at 120 kV. A Gatan Ultrascan 4000 4 k × 4 k CCD Camera System Model 895 was used to acquire the images and an EDAX Octane T Ultra W/ApolloXLT2 SDD and TEAM EDS Analysis System for EDS spectral acquisition.

Chemical analysis of aerosols

Aerosol samples were collected on 19 June 2018 using a Micro Orifice Uniform Deposit Impactor (MOUDI). The instrument was operated for 12 h runs from 6:00 am with a flow rate of 30 L/min. The instrument uses 8 impaction stages to collect size-fractionated aerosol. The concentration of metals in aerosol samples collected near the airport was obtained by ICP/MS/MS. We used a triple quadrupole 8900 Agilent and a SPS 4 auto-sampler, a quartz spray chamber, and a glass concentric nebulizer with a Nickel (Ni) interface cone and the Agilent ICP-(QQQ)-MS/MS MassHunter software for data acquisition. The analysis of metals was done in the He mode with “Low matrix” plasma condition, and auto-tuned lens voltage before each run. The standards used for the analysis were as follows: Agilent particles 8500–6940, High-Purity Standards ICP-(QQQ)-MS/MS-B with concentrations of 10 µg/cm³. Calibration was done with 2.0% HNO₃ (grade Omnitrace Ultra, EMD) in the range of 0.0–50 mg/m³. The metal data provided in Table 3 is indeed an average of the absolute concentrations obtained by subtracting a blank filter (without sample) from the sample filter. Triplicate analyses were performed for all samples. We used triplicates at each site. We also used the travelling blanks (Milli-Q water). Furthermore, three filter blanks (for aerosol analysis) have been used for each site and were analyzed three times by ICP-MS.

NAPS site description and instrumentation

Long-term continuous measurement data for NOₓ (NO + NO₂), CO, O₃, and PM₂.₅ were obtained from the Canadian National Air Pollution Surveillance Program (NAPS) stations from 2008 to 2020. The data for Montreal Airport was obtained from the NAPS station Dorval (Id:50,128, Montreal Airport, coordinates: 45.468322, −73.741198), while downtown data was collected from station Drummond downtown Montreal (Id:50,105, coordinates: 45.497859, −73.573035). The data for Vancouver airport was retrieved from Templeton street (Id:100,134, coordinates: 49.186265, −123.152422), while downtown Vancouver data was obtained from 11th avenue street downtown Vancouver (Id:100,141, coordinates: 49.260267, −123.077776). The data for Toronto airport was obtained from Jhon street (Id:60,405, coordinates: 43.636390, −79.392500), and downtown Toronto data was collected from resources road (Id:60,430, coordinates: 43.709444, −79.543500). Monthly data was obtained from the City of Montreal Air Quality website. We used the concentrations of PM₂.₅, CO, NO, NO₂, and O₃ measured at the airport of Montreal and its downtown area.

Ozone measurements were performed using a UV-photometric Analyzer (model 49i). Nitrogen oxides’ [NOₓ = (NO + NO₂)] chemiluminescence was evaluated by a NO-NO₂-NOₓ Analyzer (Model 42i). Carbon monoxide (CO) was monitored by a trace level enhanced gas filter correlation analyzer (Model 48i-TLE). PM₂.₅ was analyzed using synchronized hybrid ambient real-time analyzers (SHARP 5030) along with the PM2.5 particulates using a Met One SuperSASS-Plus Sequential Speciation Sampler. All equipment was obtained from Thermo Fisher Scientific (Forge Parkway Franklin).

Statistical analysis

In this study, the arithmetic means, standard deviation, standard error, median, 75th, and 25th percentile were calculated for data interpretation. The Spearman correlation coefficients were obtained to see if there was any correlation between variables (downtown vs airport) at each city. We also calculated the Pearson correlation coefficient to verify
Seasonal daily mean concentration of a PM$_{2.5}$ and b CO in Toronto’s airport, c PM$_{2.5}$ and d CO in Vancouver’s airport, and e PM$_{2.5}$ and f CO in Montreal’s airport in spring, summer, fall, and winter; and hourly concentration of CO and PM$_{2.5}$ at g Toronto’s airport and h downtown, i Vancouver’s airport and j downtown, and k Montreal’s airport and l downtown. Each data point represents the geometric mean at each hour of the day from January 2008 to December 2019. The error bars represent the standard error calculated as the standard deviation divided by the square root of the total number of samples for each data point.

The statistical difference of particle size distribution during the lockdown and the year prior. In Montreal, lockdown due to the COVID19 outbreak started on March 13th, and it lasted for about 4 months. On June 22nd, Montreal started partially reopening.

**Back-trajectory analysis**

To calculate the back trajectories, we used the NOAA HYSPLIT Trajectory Model. The model was run for 25 October 2019 (before COVID lockdown) and 14 April 2020 (after COVID lockdown) at the sampling location near the airport of Montreal, 45.4562° N, 73.7647° W. Each trajectory is for an interval of 24 h.

**Materials**

All chemicals used for experiments and analyses were purchased from Sigma-Aldrich. The pre-cleaned amber glass jars used for snow sampling were purchased from VWR International. Captiva Premium Syringe Filters with a 0.2 μm pore size, 25 mm wide PES (5190–5275) membrane, were obtained from Agilent Technologies.

**Results**

**Aerosol concentration near airport residential areas**

The daily PM$_{2.5}$ particle number density as well as its co-pollutants CO, NO, NO$_2$, and O$_3$ at the airports of Toronto, Vancouver, and Montreal for the four different seasons are shown in Fig. 1 and Fig. S2. The time series of these co-pollutants in the three sites are presented in the supplementary material (Fig. S3). An increase of CO, NO, NO$_2$, and PM$_{2.5}$ were observed during fall and winter, and a decrease of O$_3$ in all three airports. O$_3$ is not released directly from anthropogenic activities, unlike the other air pollutants studied in this paper, but it is produced in situ. O$_3$ is constantly being formed and destroyed when the photolysis of NO$_2$ produces excited atomic oxygen that reacts with molecular oxygen (O$_2$) (Crutzen et al. 1999). The remaining NO can react with the newly formed O$_3$ to regenerate NO$_2$. Other species, however, can regenerate NO$_2$ leading to the accumulation of O$_3$, especially in polluted areas (Edwards et al. 2013). O$_3$ can also undergo photolysis and react with water vapor to produce OH radicals and O$_2$. However, the loss of O$_3$ via these reactions is decreased during winter due to the shorter duration of sunlight and reduced water vapor concentrations (Spivakovskiy et al. 2000). In the summertime, the concentration of O$_3$ also peaks due to photochemical transformation. O$_3$ generally increases during spring and summer while decreases in fall and winter. Similar results have been reported previously (Rahim et al. 2019), where the change of O$_3$ is inversely proportional to the change of PM$_{2.5}$ and CO, NO, and NO$_2$.

Overall, the daily mean concentrations of PM$_{2.5}$ for the 10 years, at the three airports from warmest to coldest, are $5.19 \pm 0.017$ μg/m$^3$ in Vancouver, $6.71 \pm 0.019$ μg/m$^3$ in Toronto, and $7.34 \pm 0.023$ μg/m$^3$ in Montreal. As depicted in Fig. S4, the highest daily average PM$_{2.5}$ concentration was observed in Montreal, even though its airport is the smallest of the three studied airports in Canada, both in terms of flights and number of passengers (Fig. S5). Montreal is also the coldest city, which also receives the most snow each year. Toronto, the largest airport, in terms of travelers and number of flights, and being the second coldest city, has the second largest values of PM$_{2.5}$ daily averages. Vancouver airport which is the second largest airport and is in a coastal mild climate region exhibited the lowest observed average daily PM$_{2.5}$ values. A selected airborne PM$_{2.5}$ emission rate and airborne particle number density in selected airports around the world are given in Tables 1 and 2 for comparison. Airports in Canada are not amongst the airports with the highest airplane traffic and passengers such as Atlanta, Beijing, or Los Angeles; therefore, the PM$_{2.5}$ average concentrations are expected to be lower than the world’s largest airports. Yet, as shown in Tables 1 and 2, the average concentration of PM$_{2.5}$ in busy Canadian airports is similar in magnitude to airports like London Heathrow. Despite the difference in the number of passengers (at least Montreal which is more than half), the concentration is barely double. The average daily concentration trends of PM$_{2.5}$ and CO in the airport of Montreal being higher than Vancouver and Toronto can be explained through the inversion effect of the boundary layers. In cold regions during cold seasons, the bottom layer of the troposphere becomes colder than the upper layers facilitating the entrapment of gas pollutants including particles (Pal et al. 2020). This trend prevails throughout the wintertime (Fig. 1). The inversion effect is stronger in Montreal because Montreal (208 cm average yearly snow) (Canada 2021a) is colder and snowier than Toronto (112 cm average yearly snow) (Canada 2021b), and Vancouver (mild climate city with very limited snow, annual average: 34 cm) (Canada 2021c).
The increase of PM$_{2.5}$, CO, NO, and NO$_2$ during fall and winter is also in part due to the increase of idling of planes (Bern 2019), and the condensation of nanoparticles into larger particles during the cold seasons. As described previously, in cold seasons, cold temperatures alter the air density, and the temperature profile at relatively low altitudes inverts, which result in the accumulation of pollutants near surfaces (Pal et al. 2020). In warmer seasons, the height of the temperature profile at the boundary layer is increased (Argentini et al. 2005), allowing for dilution of air pollutants.

As shown in Fig. 1g–l, an increase of concentration of particles during rush hours (5:00 to 9:00 and 16:00 to 20:00 h) was consistently observed throughout the year regardless of the season with minor fluctuations the rest of the day. The concentration of CO and PM$_{2.5}$ increased during rush hours at the Montreal airport, confirming previous studies (Pal et al. 2020; Rahim et al. 2019). We observed similar trends at rush hours and non-rush hours in all other airports while downtown Vancouver shows otherwise. This is due to the chosen three cosmopolitan cities with diverse meteorology, topography, and climatology. For instance, Montreal is an island, and its meteorology and topography are different from Toronto and Vancouver (Razy et al. 2012). The primary sources in Montreal are vehicle exhaust, local industries, local heating during wintertime, and emissions from restaurants. These sources significantly contributed to PM$_{2.5}$ and CO concentrations (Pal et al. 2020).

PM$_{2.5}$ and CO are primarily emitted in the urban atmosphere by traffic emission; thus, both pollutants show strong diurnal dependency in the Montreal region. Similar trends were obtained in the

| Airport                                   | Total passengers in 2019 | Airport PM$_{2.5}$ concentration | Airport particle number density | Reference                           |
|-------------------------------------------|--------------------------|----------------------------------|--------------------------------|-------------------------------------|
| Montréal-Pierre Elliott Trudeau International Airport | 20.3 million            | 7.52 ± 0.023 μg m$^{-3}$         | 2.0 × 10$^6$ cm$^{-3}$         | (Rahim et al. 2019)                 |
| Vancouver International Airport           | 26.4 million             | 5.31 ± 0.017 μg m$^{-3}$         | N/A                            | This study                          |
| Toronto Pearson International Airport      | 50.5 million             | 6.71 ± 0.199 μg m$^{-3}$         | N/A                            | This study                          |
| Hartsfield-Jackson Atlanta International Airport | 110,531,300             | 25 μg m$^{-3}$ 70 tons year$^{-1}$ | 2.0 × 10$^4$ cm$^{-3}$         | (Riley et al. 2016; Unal et al. 2005) |
| Beijing Capital International Airport      | 100,011,438             | 69.35–119.64 μg m$^{-3}$ 149 tons year$^{-1}$ | N/A                            | (Zheng et al. 2005)                 |
| Los Angeles International Airport          | 88,068,013              | 33.03 ± 0.15 μg m$^{-3}$         | 5.0 × 10$^4$ cm$^{-3}$ (3.9–6.3) × 10$^4$ cm$^{-3}$ > 10$^7$ cm$^{-3}$ (takeoff) | (Hudda and Fruin 2016, Shirimohammadi et al. 2017; Westerdahl et al. 2008; Zhu et al. 2011) |
| Indira Gandhi International Airport        | 67,301,016              | 198.6 ± 55.6 μg m$^{-3}$         | N/A                            | (Ali et al. 2019)                   |
| London Heathrow Airport                    | 80,888,305              | 11–15 μg m$^{-3}$               | N/A                            | (Masiol and Harrison 2015)          |
| Shanghai Pudong International Airport      | 76,153,455              | 72.2 tons year$^{-1}$           | N/A                            | (Xu et al. 2020)                    |
| Santa Monica Airport                       | 88,068,013              | 18 μg m$^{-3}$ (1.5–10.6) × 10$^4$ cm$^{-3}$ | 1.98 × 10$^{24}$ year$^{-1}$ (annual PN) | (Hu et al. 2009)                    |
| Brisbane Airport                           | 24,114,833              | 1.35 × 10$^4$ kg year$^{-1}$    | 1.4–4.2) × 10$^4$ cm$^{-3}$ 3.5 × 10$^4$ cm$^{-3}$ | (Mazaheri et al. 2011)              |
| Amsterdam Airport Schiphol                 | 71,680,000              | N/A                             | 1.4 × 10$^4$ cm$^{-3}$          | (Keuken et al. 2015; Pirhadi et al. 2020) |
| Copenhagen Airport Kastrup                  | 30,300,000              | 13–17 μg m$^{-3}$ (0.12–3.3) × 10$^5$ cm$^{-3}$ | 3.5 × 10$^4$ cm$^{-3}$          | (Ellermann et al. 2012)             |
| Zurich Airport                             | 31,507,692              | N/A                             | 1.04 × 10$^5$ cm$^{-3}$         | (Fleuti et al. 2017)                |
| Heathrow Airport                           | 80,800,000              | ~ 15 μg m$^{-3}$                | 1.9 × 10$^4$ cm$^{-3}$ (Masiol et al. 2017) | (Masiol et al. 2017)                |
| Brussels Airport                           | 26,400,000              | N/A                             | (0.4–3) × 10$^5$ cm$^{-3}$      | (Stacey 2018)                       |
| Venice Airport                              | 11,561,594              | 16 μg m$^{-3}$                  | 1.4 × 10$^4$                   | (Masiol et al. 2016)                |
| Boston Airport                              | 42,522,411              | N/A                             | 1.9 × 10$^4$                   | (Hudda et al. 2018)                 |
Toronto area with a weak dependence on each other. It could be understood that major industries are located in Toronto other than the local traffic sources (Jeong et al. 2020; Tsai et al. 2004).

Downtown Vancouver shows distinct trends between PM$_{2.5}$ and CO concentrations, specifically from afternoon to early morning. It could be explained due to the different emission sources and geography of the city of Vancouver. The major sources of air pollution, including PM$_{2.5}$ and CO in Vancouver, include industrial emissions, vehicle emissions and wildfire smoke sources, local traffic, construction, and restaurants. It is also noted that Vancouver is a coastal city, and because of that, sea salt aerosol has a significant contribution to PM$_{2.5}$ along with ship traffic-borne aerosols. In contrast to Montreal and Toronto, these are the two important factors in downtown Vancouver that could cause disagreement with the diurnal variability between PM$_{2.5}$ and CO concentration (Geoff Doerksen et al. 2020).

We noted some similarities (CO and PM$_{2.5}$ increase together at the same hours in the three cities) in the concentration trends of PM$_{2.5}$ and CO in the airport of the three studies cities (Tables 1 and 2), even though the airport of Montreal has lower traffic (in terms of number of passengers) than the other two airports (Fig. S5). In short, the

### Table 2

| Metric (µg m$^{-3}$) | Montreal Airport | Montreal Downtown | Vancouver Airport | Vancouver Downtown | Toronto Airport | Toronto Downtown |
|----------------------|------------------|-------------------|-------------------|-------------------|----------------|-----------------|
| Mean                 | 7.52             | 10.6              | 5.31              | 9.42              | 6.71           | 6.33            |
| STD                  | 7.31             | 8.56              | 5.59              | 9.01              | 59.9           | 5.99            |
| Median               | 5.86             | 4.80              | 4.10              | 6.60              | 6.00           | 5.00            |
| 25%                  | 2.60             | 4.75              | 2.30              | 1.80              | 3.00           | 2.00            |
| 75%                  | 10.4             | 14.0              | 6.70              | 15.0              | 10.0           | 9.00            |
| Pearson’s correlation | 0.569            | p<0.050           | 0.0409            | p<0.050           | 0.0738         | p<0.050         |

| Metric (ppm)         | Montreal Airport | Montreal Downtown | Vancouver Airport | Vancouver Downtown | Toronto Airport | Toronto Downtown |
|----------------------|------------------|-------------------|-------------------|-------------------|----------------|-----------------|
| Mean                 | 0.17             | 0.26              | 0.26              | 0.28              | 3.29           | 0.0235          |
| STD                  | 0.12             | 0.14              | 0.17              | 0.13              | 59.1           | 0.0667          |
| Median               | 0.17             | 0.23              | 0.21              | 0.25              | 0.22           | 0               |
| 25%                  | 0.11             | 0.17              | 0.16              | 0.20              | 0.22           | 0               |
| 75%                  | 0.22             | 0.33              | 0.29              | 0.32              | 0.17           | 0               |
| Pearson’s correlation | 0.294            | p<0.050           | 0.402             | p<0.050           | 0.00463        | p= 0.164        |

| Metric (ppm)         | Montreal Airport | Montreal Downtown | Vancouver Airport | Vancouver Downtown | Toronto Airport | Toronto Downtown |
|----------------------|------------------|-------------------|-------------------|-------------------|----------------|-----------------|
| Mean                 | 3.64             | 8.92              | 11.4              | 18.8              | 7.54           | 3.11            |
| STD                  | 11.0             | 13.0              | 24.6              | 23.5              | 63.3           | 6.50            |
| Median               | 0.69             | 4.80              | 2.50              | 10.4              | 4.00           | 1.00            |
| 25%                  | 0.038            | 1.82              | 0.60              | 4.60              | 1.00           | 0               |
| 75%                  | 2.49             | 11.1              | 9.40              | 23.5              | 12.0           | 3.00            |
| Pearson’s correlation | 0.443            | p<0.050           | 0.593             | p<0.050           | 0.192          | p<0.050         |

| Metric (ppm)         | Montreal Airport | Montreal Downtown | Vancouver Airport | Vancouver Downtown | Toronto Airport | Toronto Downtown |
|----------------------|------------------|-------------------|-------------------|-------------------|----------------|-----------------|
| Mean                 | 8.94             | 15.2              | 14.6              | 19.2              | 14.1           | 12.3            |
| STD                  | 9.03             | 8.76              | 9.50              | 7.70              | 61.2           | 9.22            |
| Median               | 6.15             | 13.6              | 12.7              | 18.8              | 15.0           | 11.0            |
| 25%                  | 2.60             | 8.68              | 6.70              | 13.4              | 9.00           | 6.00            |
| 75%                  | 12.48            | 20.02             | 21.0              | 24.4              | 24.0           | 17.0            |
| Pearson’s correlation | 0.548            | p<0.050           | 0.559             | p<0.050           | 0.135          | p<0.050         |

| Metric (ppm)         | Montreal Airport | Montreal Downtown | Vancouver Airport | Vancouver Downtown | Toronto Airport | Toronto Downtown |
|----------------------|------------------|-------------------|-------------------|-------------------|----------------|-----------------|
| Mean                 | 21.9             | 19.8              | 16.7              | 9.42              | 17.3           | 22.3            |
| STD                  | 13.7             | 11.1              | 12.2              | 9.01              | 62.1           | 14.8            |
| Median               | 0.0424           | 0.0346            | 0.0382            | 0.0309            | 0.207          | 0.0413          |
| 25%                  | 22.7             | 19.2              | 16.0              | 6.60              | 20.0           | 23.0            |
| 75%                  | 31.5             | 19.2              | 5.10              | 1.80              | 30.0           | 12.0            |
| Pearson’s correlation | 0.703            | p<0.050           | 0.767             | p<0.050           | 0.194          | p<0.050         |
airport of Montreal has a higher concentration of PM$_{2.5}$ and co-pollutants than Toronto and Vancouver despite being a smaller airport with lower traffic (passengers and flights) due to its colder weather.

Statistically, PM$_{2.5}$ is higher in the downtown areas than near the airport; however, PM$_{2.5}$ is not representative of nanoparticles. Rahim et al. (2019) demonstrated that in Montreal the concentration of nanoparticles inside and near the airport is higher than that of downtown. Keep in mind that most of the mass of PM$_{2.5}$ is related to the mass of the larger particles (Zhao et al. 2020a) and the contribution of nanoparticles is generally small, and the exact contribution and particle size distribution vary depending on different factors like location, local weather conditions, and prevailing particle sources (Rönkkö and Timonen 2019). Rahim et al. (2019) found that the size fraction 180–320 nm was the most abundant. Nanoparticles are more toxic than large particles (Nemmar et al. 2013) in air, so even though the concentration of PM$_{2.5}$ is larger in downtown, air pollution due to particles is more significant near the airport. Additionally, in downtown, most particles are likely coming from road-related emissions, whereas particles near the airport come from aircraft-related emissions, since a recent study showed that road emissions and aircraft-related emissions can be distinguished by their particle size distribution and black carbon concentrations (Austin et al. 2021). Aircraft-related emissions have a relatively smaller fine particle size distribution while road-related emissions have a relatively larger fine particle size distribution (Austin et al. 2021). In terms of black carbon, it has been shown that aviation-related emissions have higher concentrations than road emissions (Li and Ariya 2021).

The Pearson correlation coefficient between downtown and the airport (0.569) is small, suggesting that they are not linearly correlated possibly because the source is not the same. The rest of the co-pollutants have a similar tendency where the downtown area has a higher concentration than the airport. The concentration of CO seems to be the most similar between the downtown areas and the airports. Still, the Pearson correlation coefficients are low (Tables 1 and 2).

**Physicochemical analysis of particles in fresh snow**

Montreal receives 2.1 m of snow over 5 to 6 months per year. The high-resolution electron microscopy imaging was used to obtain the snow particles morphology and elemental composition determination (using EDS). Using this methodology, we demonstrated that some of the airborne particulate pollutants that have been observed in the air were deposited in snow, namely various metal-containing particles (Ariya et al. 2018).

Snow has been shown to host various types of particles of both organic and inorganic nature, as well as biological particles (Mortazavi et al. 2014, 2019). It has also been shown that the existence of particles can change snow albedo or reflectivity, which directly affects the radiation budget and thus climate change (Conway et al. 1996, Warren and Wiscombe 1980). In this study, we measured fresh falling snow, and surface snow during the snow precipitation, as well as aged snow, with methodology explained elsewhere (Kos and Ariya, 2010, Kos et al. 2014). For instance, high-resolution STEM images and corresponding EDS spectra of fresh falling snow are depicted in Fig. 2. Agglomeration of small particles is shown in Fig. 2a. Both rods and tube-like particles were also observed. Figure 2b illustrates the presence of nanostructures or nanotube-like materials, which have been detected previously in highly polluted areas like oil sand regions (Rangel-Alvarado et al. 2019).

EDS elemental analysis revealed that these particles are mostly composed of carbon, hinting at combustion products (Dikio 2011). Cu peaks are related to TEM grids that are made of copper. Figure 2d and e depict particles collected directly from air. In addition to other types of nano-size particles as shown in Fig. 2, larger micron size particles were also detected (Fig. 2g). The presence of these particles suggests the burning of fossil fuels by engines as carbonaceous nanoparticles are generally produced after such activities (D’Anna 2009). There is evidence showing that nanoparticles are produced by aircraft engines; some are soot particles with graphite-type layers (Moulin et al. 2008) and also iron-bearing nanoparticles originating from impurities in the kerosene fuel used in aircrafts (Gonet and Maher 2019). This is something that does not occur with the combustion of regular gasoline fuel by vehicles.

Particles in Fig. 2 had a more complex composition containing different metals such as Fe, Mn, Mg, and Ti (Fig. 2h and i). Chlorine was also present in some particles (Fig. 2i). Chlorine salts are widely used as de-icing in cold climate cities (Hall et al. 2020). The anthropogenic chlorine is suggested to impact the oxidation potential of the atmosphere in cold urban environments (Baudrimont et al. 2020). It is worth noting that such elements have been previously observed near airports (Pal et al. 2020). Other studies of soils near airports have also revealed the presence of different metals and heavy metals and while the concentrations vary from study to study, all agree that the concentration of these species is elevated (Özkan et al. 2017; Rao et al. 2014; Ray et al. 2012). Since the EDS analysis detection is not sufficient for ultra-trace analysis, we also performed a complementary QQQ-ICP-MS/MS analysis. Table 3 presents various metal analyses in aerosols using ICP-MS/MS. As noted in Fig. 2 and Table 3, ICP-MS/MS analysis confirmed the presence of metals such as Fe, Ti, and Mg which were also detected by EDS.

Triple quad ICP/MS/MS analysis of the aerosols collected near the airport of Montreal revealed a wide range
of trace metals, including selected emerging contaminants such as Fe, Ni, Co, and Cr (Stuart and Compton 2015). The reported elevated concentrations of metals in aerosols are in accordance with aerosol measurements, which were observed both in nanoparticles as well as those larger than a micron (Rahim et al. 2019). As shown in Table 3, Al, Fe, and Zn were the most abundant metals observed in aerosols. The concentrations of several transitions and post-transition metals (e.g., Al, Fe, and Zn) were 4.04 up to 28.5 µg/L in the case of Fe. Furthermore, the concentration of metals is not consistently higher at a specific size fraction, for example, particles in the size fraction of 5.6 µm are highest for Ba, Fe, and Al, but not for Zn and other metals. However, fine particles produced by the combustion of fuels are generally enriched by trace metals which are toxic at a high concentration or prolonged exposure (Nelson 2007).

The ICP-MS analysis of particles in the snow has been shown in the airport and the reference site (downtown Montreal, McGill University). We observed that several pollutants are much higher than an already polluted site of downtown Montreal, hinting that the airports are excellent hotspots for the emission of airborne particles found in the air and deposited in snow as shown in the SI (Table S3).

The impact of COVID on airport pollution

In 2020, due to the COVID-19 pandemic, global air traffic has seen a drastic decrease (Statista 2021c), resulting in decreased aircraft-related emissions (Berman and Ebisu 2020). Several studies have reported a decrease in air pollutants such as NO2, CO, SO2, and PM2.5, and an increase in O3 in different cities around the world (Briz-Redón et al. 2020, Chauhan and Singh 2020, Kumar et al. 2020, Le et al. 2020, Mishra et al. 2021, Sicard et al. 2020, Siciliano et al. 2020, Zhao et al. 2020b). Figure 3 shows the particle concentration in the airport of Montreal in late 2019 (fall and winter) and early 2020 (winter and spring) before the lockdown was in effect vs the particle concentration during the lockdown period.

During the lockdown period, we observed a drastic decrease in the concentration of particles in the range of 0.01–1.0 µm near the airport of Montreal (Fig. 3a), and of PM2.5 (Fig. 3b). This is likely due to the decrease in the number of flights during this period. The concentration of particles before the pandemic in the airport of Montreal has been shown to peak twice a day, between 6:00 and 10:00 am and 4:00 and 8:00 pm (Rahim et al. 2019). During the lockdown period, however, we were unable to
measure the air-pollutant peaks like previously observed (Rahim et al. 2019), likely due to the sharp decrease in air traffics. Note that hourly, seasonal, and yearly variability in the concentration of submicron particles has been previously observed (Pal et al. 2020; Rahim et al. 2019). As shown in Fig. 3a, there is a significant decrease in condensed particles (0.01–1.0 μm) during the lockdown. We performed statistical analysis (the Pearson correlation coefficient of $-0.259$ and $T$-test with a $p$-value $<0.001$) indicating the distribution curves prior to and after lockdown are statistically distinct. Note the observational data were taken using a CPC counter on 5 days in October, and 5 days in April at the exact same location close to the Montreal airport. It is particularly interesting, as in cold climates during the winter, fine particles have a large peak in winter, in addition to summer photochemical peaks (Pal et al. 2020). Yet, during the lockdown, a pronounced decrease in the concentrations of PM$_{2.5}$ and submicron, including nanoparticles, in residential areas close to airports was observed ($\sim 1 \times 10^4$ cm$^{-3}$), while before the lockdown, condensable particles were up to $\sim 1 \times 10^5$ cm$^{-3}$. Hence, during the lockdown, the highest peak for condensed particles decreased by an order of magnitude, in complete contrast to previous years (Pal et al. 2020).

As illustrated in Fig. 3a, the concentration of particles during the COVID-19 lockdown was below the threshold for health exposure (2 × 10$^4$ cm$^{-3}$) for submicron particles, whereas before COVID-19, it was well above the limits. It is interesting that the PM$_{2.5}$ trend showed a significant decrease in daily airport peaks of rush hours (both morning and evening, Fig. 3b). Like Fig. 3a, the statistical analysis (the Pearson correlation coefficient of 0.0724 and $T$-test with a $p$-value $<0.001$) indicated that the distribution curves prior to and after lockdown are statistically distinct. The difference in Fig. 3b is not as large as Fig. 3a; while the difference in Fig. 3a was at least an order of magnitude, in Fig. 3b, the concentration of PM$_{2.5}$ before and after the lockdown is in the same order of magnitude. Nevertheless, the concentration of PM$_{2.5}$ is lower during the lockdown period similar to other cities (Chauhan and Singh 2020).

The exposure to jet engine emissions is an increasing health concern (Bendtsen et al. 2021; Stettler et al. 2011). Passengers and airport staff are at risk of exposure to these

### Table 3

Concentration (μg/L) of selected metals in aerosols collected near the airport of Montreal. Each color represents different elemental groups in the periodic table (white: alkali metal; green: transition metals; blue: post-transition metals; and yellow: metalloids). The metal is an absolute concentration obtained by subtracting a blank filter (without sample) from the sampled filter.

| Unit | Size fraction | 0.18 μm | 0.32 μm | 0.56 μm | 1.0 μm | 1.8 μm | 3.2 μm | 5.6 μm | 10 μm | 18 μm | LOD |
|------|---------------|--------|--------|--------|--------|--------|--------|--------|-------|-------|-----|
| **Alkaline earth metal** | **Ba** | 0.03 | 0.1 | 0.21 | 0.25 | 0.51 | 1.01 | 0.88 | 0.46 | 0.14 | 0.03 |
| **Transition metal** | **Cr** | 0.28 | 0.2 | 0.05 | 0.11 | 0.42 | 0.26 | 0.17 | 0.18 | 0.36 | 0.03 |
| | **Mn** | 0.17 | 0.11 | 0.11 | 0.11 | 0.21 | 0.44 | 0.5 | 0.45 | 0.24 | 0.03 |
| | **Fe** | 28.31 | 4.04 | 5.01 | 6.12 | 13.98 | 26.15 | 28.51 | 22.83 | 8.91 | 0.04 |
| | **Co** | 0.02 | 0.01 | 0.03 | 0 | 0 | 0.01 | 0.01 | 0.02 | 0 | 0.04 |
| | **Ni** | 0.03 | 1.12 | 0.65 | 0.1 | 0.07 | 0.68 | 0.23 | 0.25 | 0.17 | 0.03 |
| | **Cu** | 0 | 0.1 | 0.12 | 0.16 | 0.36 | 0.78 | 0.64 | 0.38 | 0.07 | 0.04 |
| | **Zn** | 26.87 | 16.65 | 1.04 | 3.14 | 0.76 | 4.22 | 1.13 | 1.24 | 0.52 | 0.05 |
| | **Cd** | 0 | 0.02 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0.03 |
| | **Mo** | 0.04 | 0.05 | 0.01 | 0.02 | 0.03 | 0.03 | 0.01 | 0.02 | 0.02 | 0.03 |
| **Post transition metal** | **Al** | 0.9 | 4.18 | 4.04 | 2.94 | 5.84 | 11.22 | 12.68 | 10.73 | 4.04 | 0.37 |
| | **Pb** | 0.03 | 0.13 | 0.03 | 0.02 | 0.03 | 0.05 | 0.05 | 0.05 | 0.04 | 0.03 |
| **Metalloids** | **As** | 0.02 | 0.02 | 0.01 | 0 | 0 | 0.01 | 0.01 | 0.01 | 0 | 0.03 |
| | **Sb** | 0.01 | 0.01 | 0 | 0 | 0.02 | 0.07 | 0.05 | 0.01 | 0.02 | 0.03 |

*ND is for concentrations under the detection limit*
emissions especially particulate matter. In many airports, the concentration of PM$_{2.5}$ can easily reach levels harmful to human health (Bendtsen et al. 2021); for example, Fig. S6 shows the PM$_{2.5}$ concentration in Toronto, Vancouver, and Montreal in 2019 vs 2020. The WHO recommends an annual mean exposure limit to PM$_{2.5}$ of 10 $\mu$g/m$^3$ and a 24-h mean exposure of 25 $\mu$g/m$^3$ (WHO 2018). The exposure limit to airborne particles varies; for example, in Germany, the recommended workplace health threshold is $2 \times 10^4$ particles/cm$^3$ (Dunn et al. 2020; IFA 2012; Van Broekhuizen et al. 2012). Figure 3a illustrates that the particle concentration during the lockdown drops to $2 \times 10^4$ particles/cm$^3$. This concentration has not been reached at least in the past 5 years (Rahim et al. 2019). The drop in concentration due to COVID-19 reveals how much pollution is generated at the airport during normal activities and how much pollution workers and residents of the area are exposed to, especially during cold seasons.

**Importance of micrometeorology, topography, and cold climate**

We performed trajectory analysis for further understanding of meteorology prior to and during the COVID-19 lockdown. An example of back-trajectory analysis is given in Fig. 3c showing that most of the air masses were coming from the west and northwest. In this example, the observation of condensable submicron particles was done in a location near the airport of Montreal on the 25 October 2019 and 14 February 2020, during the sampling period. A total of 4 trajectories in a span of 48 h were calculated for both days with an interval of 24 h for each trajectory. The calculations ended at 12:00 am of the selected day.

Fig. 3 Daily average a particle number density of condensable particles near Montreal Airport as measured by a CPC before lockdown and during lockdown. The PM$_{2.5}$ limit of exposure of $2 \times 10^4$ particles/cm$^3$ is represented by a different color (Dunn et al. 2020; IFA 2012; Van Broekhuizen et al. 2012); b PM$_{2.5}$ near Montreal Airport before lockdown and during lockdown; and c an example of back-trajectory analysis for the observation of condensable particles as given in Fig. 3a, in a location near the airport of Montreal on the 25 October 2019 and 14 February 2020, during the sampling period. A total of 4 trajectories in a span of 48 h were calculated for both days with an interval of 24 h for each trajectory. The calculations ended at 12:00 am of the selected day.
It is possible that the particles found in the samples come from the engine exhaust of planes, as previously suggested (Rahim et al. 2019), as part of the airport of Montreal is located northwest of the sampling location and a residential area with low local traffic is located to the west (Rahim et al. 2019) and plane trajectories from/to the airport include the sampling area. Moreover, strong winds are more frequent in winter (Henry-L’Heureux et al. 2021). Thereby in winter, it results in a faster dispersion of pollutants, resulting in higher pollutant concentrations at the surface in winter than in summer (Henry-L’Heureux et al. 2021). Further research is required to better understand the interactions of particulate pollutants dynamics in cold climate.

Conclusions

The current air quality policy around the world near pollution hotspots does not consider the impact of cold temperatures or snow effects. About a third of the planet’s land and many cities receive frozen precipitation every year, and thus, it is important to consider the uniqueness of the cold climate airports. Airborne particulate matter, specifically PM$_{2.5}$ as well as nano-sized particles, is mostly emitted during the takeoff and landing of airplanes (Camero 2019; Mazaheri et al. 2011; Zhu et al. 2011), in contrast to some gaseous pollutants that are continuously emitted during the flight. Hence, airport regions are highly affected by particulate matter throughout the year. Notwithstanding that since PM$_{2.5}$ measurements are based on masses, and thus heavier particles that are a less abundant group of aerosol account for most of the mass. It is thus important to measure size aggregated number density of nano-size particles, which are by far the most abundant aerosols.

We showed that the impact of cold temperatures in the entrapment of airborne particles including emerging contaminants has potential for adverse impacts, taking place in the cold seasons such as fall and winter. We provided evidence that during cold temperature seasons some of the particulate pollutions due to airplane activities are indeed transferred to snow. Since it has been shown that snow-air–water/soil photobiochemical interactions occur (Ariya et al. 2018, 2011; Mortazavi et al. 2015; Nazarenko et al. 2017), upon snowmelt, some parts of these particles or their transformed compounds enter the air, affecting adversely the air quality in these neighborhoods.

We hence propose to:

- Implement a regulatory air quality policy for the health threshold of 2×10$^4$ cm$^{-3}$ for particles in the airport and the residential areas around it to assure the health and safety of airport workers, residents, and passengers.
- Create a targeted quality policy to reduce fine particle emission as recommended for cold/snowy regions, namely during the cold seasons.
- Provide a sustainable airport/aerospace policy to improve human health and living and working conditions around airports in cold climate regions. It is noteworthy that there are new methods to distinguish between aviation and traffic fine particle emissions (Austin et al., 2021).
- Reduce airport particle emission during de-icing procedures in cold climate airports by employing sustainable technological advances, and the reduction of the usage of more sustainable fuels for airplanes.

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Author contribution R. R.: formal analysis; investigation; writing—original draft; writing—review and editing; visualization. D. P.: resources, data curation. P. A.: conceptualization; validation; writing—review and editing; visualization; supervision; project administration; funding acquisition.

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Data availability All data used for this paper will be made available upon request to the corresponding author Parisa A. Ariya (parisa.ariya@mcgill.ca). The datasets generated and/or analyzed during the current study are available in the cities’ repositories:
http://www.airqualityontario.com/history/index.php
http://ville.montreal.qc.ca/portal/page?_pageid=7237,74687650&_dad=portal&_schema=PORTAL
https://catalogue.data.gov.bc.ca/dataset/77eaeaf4-0c19-48bf-a47afa9eef01409

Declarations

Ethics approval and consent to participate We declare that we have no human participants or human data.

Consent for publication Not applicable.

Competing interests The authors declare no competing interests.

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