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The effects of urban green space and road proximity to indoor traffic-related $PM_{2.5}$, NO_2, and BC exposure in inner-city schools

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BACKGROUND: Since there are known adverse health impacts of traffic-related air pollution, while at the same time there are potential health benefits from greenness, it is important to examine more closely the impacts of these factors on indoor air quality in urban schools.

OBJECTIVE: This study investigates the association of road proximity and urban greenness to indoor traffic-related fine particulate matter ($PM_{2.5}$), nitrogen dioxide (NO_2), and black carbon (BC) in inner-city schools.

METHODS: $PM_{2.5}$, NO_2, and BC were measured indoors at 74 schools and outdoors at a central urban over a 10-year period. Seasonal urban greenness was estimated using the Normalized Difference Vegetation Index (NDVI) with 270 and 1230 m buffers. The associations between indoor traffic-related air pollution and road proximity and greenness were investigated with mixed-effects models.

RESULTS: The analysis showed linear decays of indoor traffic-related $PM_{2.5}$, NO_2, and BC by 60%, 35%, and 22%, respectively for schools located at a greater distance from major roads. The results further showed that surrounding school greenness at 270 m buffer was significantly associated ($p < 0.05$) with lower indoor traffic-related $PM_{2.5}$: −0.068 (95% CI: −0.124, −0.013), NO_2: −0.139 (95% CI: −0.185, −0.092), and BC: −0.060 (95% CI: −0.115, −0.005). These associations were stronger for surrounding greenness at a greater distance from the schools (buffer 1230 m) $PM_{2.5}$: −0.101 (95% CI: −0.156, −0.046) NO_2: −0.122 (95% CI: −0.169, −0.075) BC: −0.080 (95% CI: −0.136, −0.026). These inverse associations were stronger after fully adjusting for regional pollution and meteorological conditions.

IMPACT STATEMENT: More than 90% of children under the age of 15 worldwide are exposed to elevated air pollution levels exceeding the WHO's guidelines. The study investigates the impact that urban infrastructure and greenness, in particular green areas and road proximity, have on indoor exposures to traffic-related $PM_{2.5}$, NO_2, and BC in inner-city schools. By examining a 10-year period the study provides insights for air quality management, into how road proximity and greenness at different buffers from the school locations can affect indoor exposure.

Keywords: Air pollution; Child exposure/Health; Particulate matter; Environmental monitoring

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INTRODUCTION

Primary schools are the second most important indoor micro-environment for children (other than homes) representing a unique and important location. Exposures to traffic-related pollutants in school may have a substantial impact on health of children who typically are in classrooms for over 6 h/day [1]. Schools that are located in close proximity to busy roads have higher exposures to traffic-related air pollutants (TRAPs) due to daytime traffic peaks and during morning and afternoon drop off/ pick up times [2, 3]. Exposure to TRAPs at school has been associated with a range of adverse health impacts including respiratory conditions [4–6] and impaired neurodevelopment in

schoolchildren, which contribute to a considerable personal and societal burden. In the USA, traffic-related sources account for 48% of the transportation $PM_{2.5}$ and are the largest source of NO_x in an urban environment [7] accounting for 55%. There have been several studies that highlight the impact of road proximity to indoor traffic-related pollutants [8–10]. However, the quantitative aspect and the mitigation means of this association are less explored in school environments.

The implementation of green infrastructure is increasingly recognized as a promising strategy for mitigating the negative impacts of traffic pollution. There is mounting evidence that proximity of residence to city parks and regular visits to green

spaces are associated with health benefits, including physical activity, social coherence, and stress reduction pathways [11, 12]. Greenness around schools is thought to improve health through direct effects on cognitive restoration and stress reduction [13] as well as by mitigating exposure to air pollution, noise, and extreme temperatures [14–16]. Past studies have shown that greenness provides the beneficial effects of lowering levels of both outdoor [17, 18] and indoor $PM_{2.5}$ at residences [19] and schools [20]. However, these findings are inconsistent, and show strong regional heterogeneity [21, 22]. As a concept, green space represents diverse landscape features in myriad arrangements, with a variety of functionalities [13]. Consequently, the interaction between green infrastructure design (i.e., species selection, spatial positioning) and air pollutants can either positively or negatively affect personal exposure and thus human health [23].

Since traffic-related pollutants such as $PM_{2.5}$, BC, and NO_2 have been associated with adverse health effects, any amount of exposure reduction due to green space may provide benefits to children whose respiratory systems are still developing [14, 24]. In this study, conducted in a north-eastern USA city, we evaluated the long-term impacts of road proximity and quantified greenness levels in an inner-city environment, on the traffic-related air pollutant exposures inside school classrooms. This study is building upon the data from The School Inner City Asthma Study I & II (SICAS 1 & 2) where previously published results have identified and quantified the controllable factors and sources that influence outdoor and indoor air pollution exposures [2, 25]. How proximity to major roadways at home and school increase asthma symptoms [26] and how NO_2 levels inside classrooms even at low levels affect airflow obstruction [6].

METHODS
Study design
SICAS1 and SICAS2 investigated the effect of school- and classroom-based environmental exposures on students with asthma in a city in the north-eastern USA. The study spanned 10 years 2008–2014; 2015–2019 and included classroom measurements throughout the academic school. Weeklong indoor NO_2, $PM_{2.5}$, and BC measurements were conducted in inner-city school classrooms during weekdays, incorporating both occupied and non-occupied periods. SICAS1 is a prospective study evaluating the school/classroom-specific risk factors and asthma morbidity among urban children. It included 350 elementary school-aged children with asthma from multiple classrooms in 38 inner-city schools. Recruitment was ongoing for 5 years and started in spring 2008 with a follow-up sampling in autumn or winter. SICAS2 was a randomized control trial of 247 students with asthma to test a school-classroom level intervention in 41 schools, where the baseline measurements occurred between October and November, the first follow-up December–February, and the second follow-up between March and May. In total, 309 unique classrooms of unique 74 schools were studied here. The rationale of the studies is described in detail elsewhere [27, 28].

Air pollution measurements
Weekday-period indoor $PM_{2.5}$ samples were collected on Teflon filters at a flow rate of 1.8 L/min [29]. A total of 518 indoor $PM_{2.5}$ samples were collected during the study period. $PM_{2.5}$ collected on Teflon filters was measured gravimetrically with an electronic microbalance (MT-5 Mettler Toledo, Columbus, OH) in a temperature/RH-controlled room, following USEPA guidelines. Indoor filters were also measured for BC concentrations using a Smokestain Reflectometer (Model EEL M43D, Diffusion Systems Ltd., UK). Indoor NO_2 was collected using Ogawa passive samplers and analyzed using ion chromatography. Concurrent daily outdoor $PM_{2.5}$, BC, and NO_2 concentrations were also measured at a central monitoring site. $PM_{2.5}$ samples were collected using the Harvard Impactor [30], BC concentrations were measured using a single $(\lambda = 880\,nm)$ channel Aethalometer (model AE-16, Magee Scientific, Berkeley, CA), and NO_2 was measured with chemiluminescent monitors. Indoor and outdoor samples were compared by matching the weekly indoor samples to the corresponding average daily outdoor samples. The median distance between the central site and schools was 4974 m. The contribution of

traffic to $PM_{2.5}$ concentrations was estimated with receptor modeling as described elsewhere [25]. This is a $PM_{2.5}$ source that is derived by the $PM_{2.5}$ mass and elemental composition using the positive matrix factorization modeling technique developed by USEPA [31, 32]. After the source apportionment 420 out of the 518 $PM_{2.5}$ samples were found to have contributions from traffic sources. This traffic-related $PM_{2.5}$ component (referred to as T-$PM_{2.5}$) was abundant in organic and elemental carbon from motor exhaust, and metals from brake and tire wear (Fe, Cu, Zn). For traffic-related NO_2 and BC we excluded samples when the indoor:outdoor ratio of NO_2 was higher than 1.2 which is likely to indicate influence of indoor sources [33, 34].

Urban green space
We estimated the greenness around schools by the use of Normalized Difference Vegetation index (NDVI). NDVI is a satellite-derived measure that captures photosynthetic activity of vegetation [35]. NDVI values range from -1.0 to 1.0, with negative values indicating clouds, snow and water, positive values near zero indicating bare soil, and higher positive values of NDVI ranging from sparse vegetation (0.1–0.5) to dense green vegetation (0.6 and above). NDVI data for every season between 2008 and 2019 at a 30 m resolution were obtained from Google Earth Engine [36]. Landsat 7 data were used for 2008–2013 and Landsat 8 for 2014–2019. We applied Google Earth Engine's cloud cover algorithm to retain the least cloudy image within each season (January–March, April–June, July–September, October–December). NDVI within each season were assessed within 270 and 1230 m surrounding each geocoded school address, to represent viewable and walkable areas around each school. The 270 m buffer was selected to represent greenness directly accessible outside each school, while the 1230 m buffer represent a walkable distance buffer as reported in the literature [37].

Distance to roadway
The distance between each school to the nearest primary or secondary road was calculated using Arc-GIS 10.2.2 (Environmental Systems Research Institute, Redlands, CA) software with 80% spelling sensitivity and 10-meter offset. First, a state-wide school database provided by the Office of Geographic Information to geocode school locations was mapped. We used a buffer of 300 m around each address to calculate the straight-line distance to the nearest primary and secondary roads [26]. Primary and secondary roads were defined using TIGER/Line files within the federal highway system. According to the US Census Bureau, primary roads are limited-access highways within the Federal interstate highway system or under state management with interchanges and accessible by ramps, including some toll highways. Secondary roads are main arteries, usually in the U.S. highway, state highway, or county highway system, with one or more lanes of traffic in each direction, may or may not be divided, and usually have at-grade intersections with many other roads and driveways.

Statistical analysis
Following the approach of Dadvand et al. [21], we developed mixed-effects models for each pollutant, with schools included as a random effect to account for the repeated measurements within each school. We used weeklong indoor levels of each pollutant as the dependant variable and greenness in the 270 or 1230 m surrounding each school as a fixed-effect predictor. Given that schools were monitored in different weeks during each campaign period, we adjusted the analyses of each TRAP for the weekly average level of that TRAP (during the corresponding sampling week for each weeklong school indoor concentration with its corresponding from the central urban background site) measured by an urban background monitoring station to remove temporal fluctuation in background TRAP levels from our analyses [38]. Because we only had one background site when adjusting for background levels we included a random slope which represents the spatial relationship of each school to that site [2]. We further adjusted the models of indoor T-$PM_{2.5}$, BC, and NO_2 for outdoor concentrations, seasonality ($\cos d = \cos (2 \times p \times d/365)$) due to variations in greenness and differences in emissions, meteorological parameters (ambient temperature, wind speed, and boundary layer height), number of windows, and window orientation (facing toward bus drop off/pick up area), and school characteristics including building age and ventilation as fixed-effect predictors. The model is illustrated in the following equation:

$$Y_{ij} = \beta_0 + X_{ij}\beta' + \mu_i + u_{ij} + \varepsilon_{ij}$$

where Y_{ij} represents the response variable weeklong indoor concentrations in school i and week j. X_{ij} is a matrix of fixed effects predictors in school i and
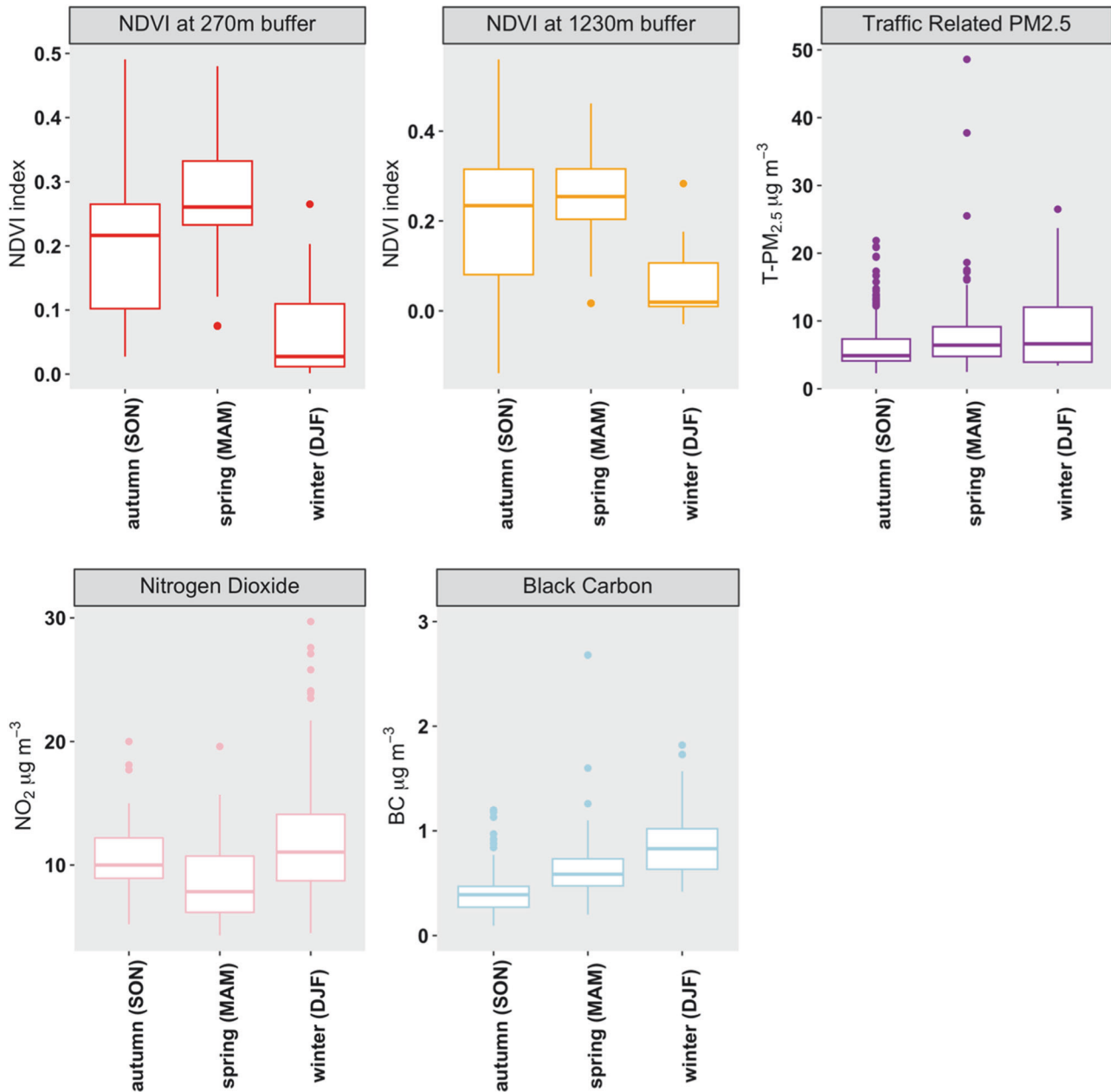
Fig. 1 **Seasonal levels of NDVI around schools and indoor T-PM$_{2.5}$ ($n = 420$), NO$_2$ ($n = 351$), and BC ($n = 372$) levels between 2008 and 2015; 2015–2019.** Box parameters are the interquartile range (IQR), the hash mark is the median, and whiskers extend to 1.5 times the IQR above the 75th and below the 25th percentiles.

week j (i.e., outdoor concentration, boundary layer height, greenness, seasonality, etc.). β' is a vector of fixed-effect coefficient. μ_i is the random intercept representing school-specific variability (between classrooms). u_{ij} is the random slope representing slope of the fixed-effect predictors that varies randomly across different schools. ε_{ij} is the error term. The inclusion of both school-specific variations (μ_i) and variations in the slopes of the fixed effects across schools (u_{ij}) can be particularly useful when the relationship between predictors and the response varies across different schools.

RESULTS
The median greenness around schools at 270 m between 2010 and 2019 was 0.204 (IQR = 0.163), with a mean value of 0.198 (s.d. = 0.122). It had a clear seasonal pattern with higher values during spring and lower values during winter. The mean (±s.d.) values were 0.217 (±0.107), 0.060 (±0.064), and 0.272 (±0.100) for autumn, winter, and spring, respectively (Fig. 1). Similar seasonal

variations were observed with the 1270 m buffer. Figure 1 shows the seasonal variations of greenness and indoor T-PM$_{2.5}$, NO$_2$, and BC. Mean T-PM$_{2.5}$ levels were 6.4 (±3.9), 8.8 (±5.7), and 8.4 (±6.8) for autumn, winter, and spring, respectively. Mean indoor NO$_2$ concentrations were 10.5 (±3.3), 11.9 (±4.5), and 9.0 (±3.6) while BC levels were 0.40 (±0.2), 0.9 (±0.3), and 0.7 (±0.8) for autumn, winter, and spring, respectively.

Figure 2 shows the relationship between the normalized (to the mean) indoor T-PM$_{2.5}$, NO$_2$, BC, and roadway proximity. It is evident that all traffic-related pollutants follow a linear decay with distance. The sharpest decrease with distance is for T-PM$_{2.5}$ followed by NO$_2$, while BC has a smoother decrease with distance. Schools that are more than 3 km away from roadways experience (on average) a 63%, 35%, and 22% decrease in T-PM$_{2.5}$, NO$_2$, and BC compared to those that are close to major roadways. Schools that have increased green areas directly accessible outside each
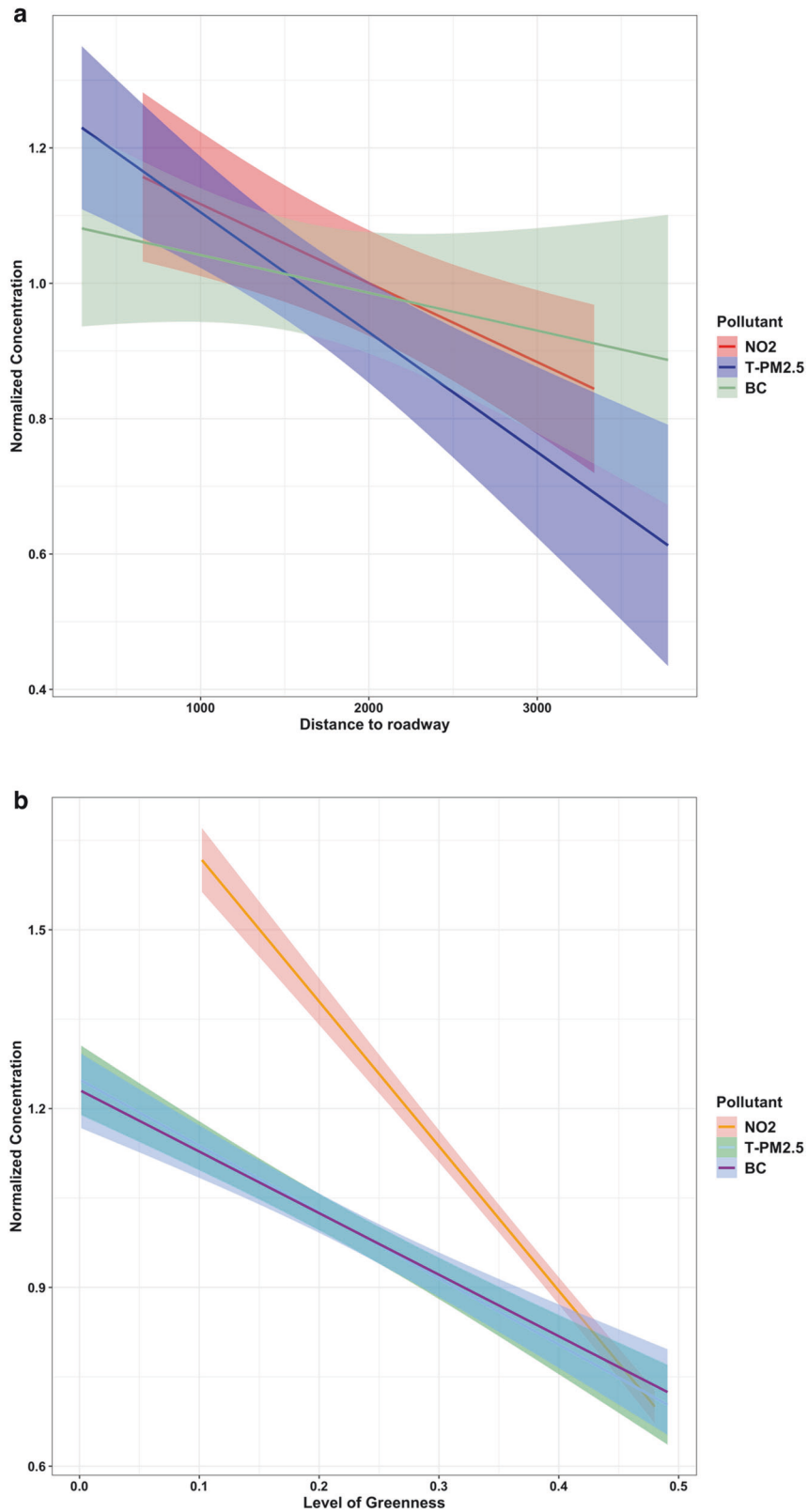
Fig. 2 Effects of green space and road proximity. a Relationship between normalised indoor traffic-related air pollution and road distance, **b** relationship between normalised indoor traffic-related air pollution and greenness at 270 m. The shaded area indicates confidence intervals at 95%. T-PM$_{2.5}$ ($n = 420$), NO$_2$ ($n = 351$), BC ($n = 372$).

Table 1. Regression coefficients and confidence intervals (CI) at 95% indicating the change of indoor T-PM$_{2.5}$, NO$_2$, and BC per NDVI increase at 270 and 1230 m of the school boundaries and road proximity.

Model	NDVI 270 m		NDVI 1230 m		Road distance	
	Coefficients (95% CI)	p	Coefficients (95% CI)	p	Coefficients (95% CI)	p
T-PM$_{2.5}$						
Unadjusted	−0.068 (−0.124, −0.013)	<0.01	−0.101 (−0.156, −0.046)	<0.01	−5.48e-05 (−1.24e-04, −1.46e-05)	0.12
Adjusted[a]	−0.109 (−0.149, −0.068)	<0.01	−0.136 (−0.174, −0.098)	<0.01	−1.00e-04 (−1.48e-04, −5.39e-05)	<0.01
NO$_2$						
Unadjusted	−0.139 (−0.185, −0.092)	<0.01	−0.122 (−0.169, −0.075)	<0.01	−1.21e-04 (−1.61e-04, −8.10e-05)	<0.01
Adjusted[a]	−0.550 (−0.843, −0.262)	<0.01	−0.361 (−0.656, −0.067)	0.016	−6.69e-05 (−1.03e-04, −3.07e-05)	<0.01
BC						
Unadjusted	−0.060 (−0.115, −0.005)	0.032	−0.080 (−0.136, −0.026)	0.004	−1.20e-05 (−8.01e-05, −5.64e-05)	0.18
Adjusted[a]	−0.069 (0.119, −0.019)	0.048	−0.103 (−0.147, −0.058)	<0.01	−8.62e-05 (−1.35e-04, −3.68e-05)	<0.01

[a]Adjusted model for regional pollution levels, wind speed, temperature, seasonality, number of windows, window orientation, and school ventilation.

school at 270 m showed great reductions in traffic-related pollutants by 64%, 61%, and 107% in T-PM$_{2.5}$, BC, and NO$_2$ compared to schools with little to no greenness. NO$_2$ showed a sharper and more abrupt decrease with increased greenness while T-PM$_{2.5}$ and BC showed similar less abrupt decreases. Similar decreases were found for the 1230 m buffer where NO$_2$ decreased by 118% while T-PM$_{2.5}$ and BC showed 68% and 67% reductions, respectively.

The results from the regression analysis of greenness and road proximity to traffic-related indoor PM$_{2.5}$, NO$_2$, and BC are shown in Table 1. For the road proximity in the unadjusted model there were inverse associations with T-PM$_{2.5}$ and BC, however, they were not significant as in the case of NO$_2$. After adjusting the models all relationships became significant ($p < 0.05$). Higher NDVI was statistically significantly ($p < 0.05$) associated with lower indoor T-PM$_{2.5}$, NO$_2$, and BC at a buffer of 270 m from schools. The coefficient of the association was stronger for greenness at a buffer of 1230 m from the schools. Analyses additionally adjusted for regional outdoor PM$_{2.5}$, NO$_2$, and BC, as well as meteorological parameters such as wind speed, temperature, boundary layer height, seasonality, and school characteristics: ventilation type and number of windows. The relationship between greenness and indoor T-PM$_{2.5}$, NO$_2$, and BC for the adjusted models had stronger coefficients for both 270 and 1230 m. Indoor T-PM$_{2.5}$, NO$_2$, and BC levels are influenced to some extent by regional pollution and the meteorological factors. Therefore, when controlling for these factors the associations between greenness and indoor T-PM$_{2.5}$, NO$_2$, and BC became stronger, allowing for a more accurate assessment of the effect of greenness and road proximity on indoor air pollution and minimizing the potential confounding effects of other variables.

DISCUSSION
The study investigated the associations of roadway proximity and greenness with exposures to traffic-related PM$_{2.5}$, NO$_2$, and BC inside inner-city schools. The analysis was based on 420, 362, and 372 PM$_{2.5}$, NO$_2$, and BC sample pairs, respectively that were collected indoors, in 74 inner-city schools, and outdoors, at a central urban background location. The results showed statistically significant declines of indoor traffic-related PM$_{2.5}$, NO$_2$, and BC with increases in roadway distance and greenness.

The decreases in T-PM$_{2.5}$ exposure with road distance were greater than 60% for schools that are located 3 km away from roadways, while NO$_2$ dropped by one third and BC by one fifth for the same schools. Amram et al. [39] showed similar findings for NO and ultrafine particles (both tracers for traffic-related pollution). These results show the importance of school location to indoor exposures to traffic-related air pollution which consequently have been shown

to affect children's lung development and health [40, 41]. Road proximity also affects indoor PM$_{2.5}$ composition and elements related to tires, brakes, and road dust have more dramatic decreases [10]. Studies have also shown that road proximity of schools is associated with increased asthma symptom days for asthmatic children [26], while it has also been associated with increased childhood leukemia [42] and neurobehavioral disorders [43]. Furthermore, road proximity in general has been shown to influence birth outcomes [44] and has been associated with increased Parkinson's disease risk [45], cognitive impairment [46], and with insidious effects on structural brain aging, even in dementia- and stroke-free persons [47].

The results further showed that surrounding school greenness was associated with a >60% reduction in traffic-related pollutants providing evidence for an urban level intervention to reduce children's exposure to traffic-related air pollution. School greenness was also significantly associated with lower indoor traffic-related pollutants. These associations were stronger for surrounding greenness at a greater distance from schools highlighting the importance of open green areas at an inner-city environment. Open green areas can offer more protective means in buffering indoor school pollution levels by gradually reducing traffic-related pollution over space. Additionally, more green areas at a greater distance could also potentially mean fewer roadways. Our results were consistent with findings of previous studies reporting inverse associations between exposures to traffic-related pollutants inside schools and surrounding greenness at different distances from the school locations [21]. Improvements between indoor PM$_{2.5}$ exposure in areas with high surrounding green space have been more consistent in the literature [48]. This link is straightforward since trees, shrubs, and hedges often act as filtration media (dry deposition) to outdoor PM$_{2.5}$ [49, 50] and BC [51], however, their type (i.e., needle leaf, broad leaf, etc.), effective density, and porosity strongly influence their effectiveness [52, 53]. Even though there have been studies reporting NO$_2$ reductions naturally onto vegetation [54, 55] or in artificial biofilters [56] via dry deposition, the reality is more complicated since more vegetation will result in greater biogenic volatile organic compound (BVOC) emissions in the near-school environment. These increased BVOC emissions are involved in complex photochemical reactions with NO$_2$ and ozone (O$_3$) and in the generation of secondary organic aerosols, therefore, might offset some of the health benefits of the overall greenness.

Previous studies that investigated factors that affect indoor school exposure to PM$_{2.5}$, NO$_2$, and BC showed that the biggest contributor to indoor air pollution is the infiltration of outdoor air pollution [2]. The association between surrounding greenness and indoor exposure to traffic-related pollution is, therefore, affected by the reductions of the outdoor traffic-related pollution with greenness.

To examine that here, we fully adjusted our models for factors such as regional pollution and meteorological variables (such as wind speed, temperature), school ventilation practices, and seasonality. We found that the association between greenness and indoor exposure becomes stronger after adjusting for these variables, suggesting that part of the benefits inside schools are likely due to the benefits of the overall reduction of outdoor traffic-related pollutants by greenness. Surrounding school greenness, not only reduces overall traffic-related air pollution, but it can also improve cognitive performance of children [57], even when children are exposed only for short time periods to nature [58]. Other benefits related to residential greenness for children include lower obesity levels [20], fetal growth [59], reduction of the risk of cardiovascular disease [60], and decreased risk of cancer mortality [61].

Despite that our results can be considered robust, spanning 74 schools over 10 years, there are several limitations in this study. The data are from one region in north-eastern US and the findings might vary for other regions depending on the city/region – specific urban infrastructure, greens species variety (i.e., trees, shrubs, plants, etc.) and vehicle fleet emissions (i.e., proportion of heavy duty vehicles vs passenger cars vs light duty vehicles, diesel vs gasoline vehicles, etc.). In addition, the study did not include outdoor sampling directly outside the schools and instead included one regional background site location and surrogates to adjust for outdoor concentrations. To ensure consistency and robustness future studies should also consider expanding this approach by including multiple cities that can represent both north and south multinational environments as well as high-middle and low-middle income countries. Due to city-wide air pollution exposure disparities future studies should also consider if and how socioeconomic factors play a role in these associations.

IMPLICATIONS FOR CHILDREN'S EXPOSURE AND RECOMMENDATIONS FOR IMPROVEMENT
According to the World Health Organization (WHO), more than 90% of children under the age of 15 worldwide are exposed to air pollution levels exceeding the WHO's recommended guidelines [62]. Given that children spend a significant portion of their day in school, it is crucial to enhance our knowledge regarding their exposure pathways to air pollution. In this large school indoor exposure study conducted in 74 schools across 10 years, the findings showed that children attending schools located near busy roads, not necessarily highways, and schools that have less nearby green infrastructure are exposed to greater traffic-related $PM_{2.5}$, NO_2, and BC and are likely at a greater health hazard risk. In a recent literature review regarding the air quality around schools, Osborne et al. [63] indicated that nearby traffic is a key determinant of concentrations outside schools and that factors related to planning and urban design such as green playgrounds, and amount of surrounding green space can reduce school site air pollution. Given the adverse health impacts of traffic-related air pollution and the improved health benefits of greenness, actions are needed to improve children's health and well-being during their early years of development. Green infrastructure, natural or artificial via botanical biofilters, in and around schools as well as the creation of clean air zones at an effective road proximity from schools are two key reduction measures that local authorities, policy makers, school managers, and urban planners should evaluate and consider as unique or combined interventions in order to reduce children's exposure to air pollution and improve their health and well-being.

DATA AVAILABILITY
The data are available from the corresponding author on reasonable request.

REFERENCES
1. Esty B, Permaul P, DeLoreto K, Baxi SN, Phipatanakul W. Asthma and allergies in the school environment. Clin Rev Allergy Immunol. 2019;57:415–26. https://doi.org/10.1007/s12016-019-08735-y.
2. Matthaios VN, Kang CM, Wolfson JM, Greco KF, Gaffin JM, Hauptman M, et al. Factors influencing classroom exposures to fine particles, black carbon, and nitrogen dioxide in inner-city schools and their implications for indoor air quality. Environ Health Perspect. 2022;130:047005.
3. Kumar P, Omidvarborna H, Pilla F, Lewin N. A primary school driven initiative to influence commuting style for dropping-off and picking-up of pupils. Sci Total Environ. 2020;727:138360.
4. McConnell R, Islam T, Shankardass K, Jerrett M, Lurmann F, Gilliland F, et al. Childhood incident asthma and traffic-related air pollution at home and school. Environ Health Perspect. 2010;118:1021–6. https://doi.org/10.1289/ehp.0901232.
5. Madureira J, Paciência I, Rufo J, Ramos E, Barros H, Teixeira JP, et al. Indoor air quality in schools and its relationship with children's respiratory symptoms. Atmos Environ. 2015;118:145–56.
6. Gaffin JM, Hauptman M, Petty CR, Sheehan WJ, Lai PS, Wolfson JM, et al. Nitrogen dioxide exposure in school classrooms of inner-city children with asthma. J Allergy Clin Immunol. 2018;141:2249–2255.e2. https://doi.org/10.1016/j.jaci.2017.08.028.
7. US EPA 2020 Environmental Protection Agency, National Emission Inventories 2014. 2018 (https://www.epa.gov/air-emissions-inventories/2014-national-emissions-inventory-nei-data#doc).
8. Janssen NA, van Vliet PH, Aarts F, Harssema H, Brunekreef B. Assessment of exposure to traffic related air pollution of children attending schools near motorways. Atmos Environ. 2001;35:3875–84.
9. Kim JJ, Smorodinsky S, Lipsett M, Singer BC, Hodgson AT, Ostro B. Traffic-related air pollution near busy roads: the East Bay Children's Respiratory Health Study. Am J Respir Crit Care Med. 2004;170:520–6.
10. Huang S, Lawrence J, Kang CM, Li J, Martins M, Vokonas P, et al. Road proximity influences indoor exposures to ambient fine particle mass and components. Environ Pollut. 2018;243:978–87.
11. Lee AC, Maheswaran R. The health benefits of urban green spaces: a review of the evidence. J Public Health. 2011;33:212–22.
12. Van den Berg M, Wendel-Vos W, van Poppel M, Kemper H, van Mechelen W, Maas J. Health benefits of green spaces in the living environment: a systematic review of epidemiological studies. Urban For Urban Green. 2015;14:806–16.
13. Hartig T, Mitchell R, De Vries S, Frumkin H. Nature and health. Annu Rev Public Health. 2014;35:207–28.
14. HEI (Health Effects Institute). Systematic review and meta-analysis of selected health effects of long-term exposure to traffic-related air pollution. Special Report 23. Boston, MA: Health Effects Institute; 2022.
15. Gidlöf-Gunnarsson A, Öhrström E. Noise and well-being in urban residential environments: the potential role of perceived availability to nearby green areas. Landsc Urban Plan. 2007;83:115–26.
16. Lafortezza R, Carrus G, Sanesi G, Davies C. Benefits and well-being perceived by people visiting green spaces in periods of heat stress. Urban For Urban Green 2009;8:97–108.
17. Nowak DJ, Crane DE, Stevens JC. Air pollution removal by urban trees and shrubs in the United States. Urban For Urban Green. 2006;4:115–23.
18. Franchini M, Mannucci PM. Mitigation of air pollution by greenness: a narrative review. Eur J Intern Med. 2018;55:1–5.
19. Ozdemir H. Mitigation impact of roadside trees on fine particle pollution. Sci Total Environ. 2019;659:1176–85.
20. Dadvand P, Villanueva CM, Font-Ribera L, Martinez D, Basagaña X, Belmonte J, et al. Risks and benefits of green spaces for children: a cross-sectional study of associations with sedentary behavior, obesity, asthma, and allergy. Environ Health Perspect. 2014;122:1329–35. https://doi.org/10.1289/ehp.1308038.
21. Dadvand P, Rivas I, Basagaña X, Alvarez-Pedrerol M, Su J, Pascual MDC, et al. The association between greenness and traffic-related air pollution at schools. Sci Total Environ. 2015;523:59–63.
22. Ferrante G, Asta F, Cilluffo G, De Sario M, Michelozzi P, La Grutta S. The effect of residential urban greenness on allergic respiratory diseases in youth: a narrative review. World Allergy Organ J. 2020;13:100096.
23. Lambert KA, Bowatte G, Tham R, Lodge C, Prendergast L, Heinrich J, et al. Residential greenness and allergic respiratory diseases in children and adolescents – a systematic review and meta-analysis. Environ Res. 2017;159:212–21.
24. Abhijith KV, Kumar P, Gallagher J, McNabola A, Baldauf R, Pilla F, et al. Air pollution abatement performances of green infrastructure in open road and built-up street canyon environments – a review. Atmos Environ. 2017;162:71–86.
25. Carrion-Matta A, Kang CM, Gaffin JM, Hauptman M, Phipatanakul W, Koutrakis P, et al. Classroom indoor $PM_{2.5}$ sources and exposures in inner-city schools. Environ Int. 2019;131:104968.

26. Hauptman M, Gaffin JM, Petty CR, Sheehan WJ, Lai PS, Coull B, et al. Proximity to major roadways and asthma symptoms in the School Inner-City Asthma Study. J Allergy Clin Immunol. 2020;145:119–26.

27. Phipatanakul W, Bailey A, Hoffman EB, Sheehan WJ, Lane JP, Baxi S, et al. The School Inner-city Asthma Study: design, methods, and lessons learned. J Asthma. 2011;48:1007–14.

28. Phipatanakul W, Koutrakis P, Coull BA, Kang CM, Wolfson JM, Ferguson ST, et al. The School Inner-cityasthma Intervention Study: design, rationale, methods, and lessonslearned. Contemp Clin Trials. 2017;60:14–23.

29. Demokritou P, Kavouras IG, Ferguson ST, Koutrakis P. Development and laboratory performance evaluation of a personal multipollutant sampler for simultaneous measurements of particulate and gaseous pollutants. Aerosol Sci Technol. 2001;35:741–52.

30. Koutrakis P, Sioutas C, Ferguson ST, Wolfson JM, Mulik JD, Burton RM. Development and evaluation of a glass honeycomb denuder filter pack system to collect atmospheric gases and particles. Environ Sci Technol. 1993;27:2497–501.

31. Norris G, Duvall R, Brown S, Bai S. EPA Positive Matrix Factorization (PMF) 5.0 Fundamentals and User Guide. Washington, DC: US Environmental Protection Agency; 2014.

32. Matthaios VN, Lawrence J, Martins MA, Ferguson ST, Wolfson JM, Harrison RM, et al. Quantifying factors affecting contributions of roadway exhaust and non-exhaust emissions to ambient $PM_{10-2.5}$ and $PM_{2.5-0.2}$ particles. Sci Total Environ. 2022;835:155368.

33. Hu Y, Zhao B. Relationship between indoor and outdoor NO_2: a review. Build Environ. 2020;180:106909.

34. Habre R, Coull B, Moshier E, Godbold J, Grunin A, Nath A, et al. Sources of indoor air pollution in New York City residences of asthmatic children. J Expo Sci Environ Epidemiol. 2014;24:269–78.

35. Kriegler F, Malila W, Nalepka R, Richardson W. Preprocessing transformations and their effects on multispectral recognition. In: Proceedings of the Sixth International Symposium on Remote Sensing of Environment. Ann Arbor, MI: University of Michigan; 1969. p. 97–131.

36. Gorelick N, Hancher M, Dixon M, Ilyushchenko S, Thau D, Moore R. Google Earth Engine: planetary-scale geospatial analysis for everyone. Remote Sens Environ. 2017;202:18–27.

37. James P, Hart JE, Banay RF, Laden F. Exposure to greenness and mortality in a nationwide prospective cohort study of women. Environ Health Perspect. 2016;124:1344–52.

38. Rivas I, Viana M, Moreno T, Pandolfi M, Amato F, Reche C, et al. Child exposure to indoor and outdoor air pollutants in schools in Barcelona, Spain. Environ Int. 2014;69:200–12.

39. Amram O, Abernethy R, Brauer M, Davies H, Allen RW. Proximity of public elementary schools to major roads in Canadian urban areas. Int J Health Geogr. 2011;10:1–11.

40. Gauderman WJ, Vora H, McConnell R, Berhane K, Gilliland F, Thomas D, et al. Effect of exposure to traffic on lung development from 10 to 18 years of age: a cohort study. Lancet. 2007;369:571–7.

41. Mohai P, Kweon BS, Lee S, Ard K. Air pollution around schools is linked to poorer student health and academic performance. Health Aff. 2011;30:741–52.

42. Houot J, Marquant F, Goujon S, Faure L, Honoré C, Roth MH, et al. Residential proximity to heavy-traffic roads, benzene exposure, and childhood leukemia—The GEOCAP Study, 2002–2007. Am J Epidemiol. 2015;182:685–93.

43. Kim SS, Vuong AM, Dietrich KN, Chen A. Proximity to traffic and exposure to polycyclic aromatic hydrocarbons in relation to attention deficit hyperactivity disorder and conduct disorder in US children. Int J Hyg Environ Health. 2021;232:113686.

44. Wilhelm M, Ritz B. Residential proximity to traffic and adverse birth outcomes in Los Angeles county, California, 1994–1996. Environ Health Perspect. 2003;111:207–16.

45. Yuchi W, Sbihi H, Davies H, Tamburic L, Brauer M. Road proximity, air pollution, noise, green space and neurologic disease incidence: a population-based cohort study. Environ Health. 2020;19:8. https://doi.org/10.1186/s12940-020-0565-4.

46. Yao Y, Jin X, Cao K, Zhao M, Zhu T, Zhang J, et al. Residential proximity to major roadways and cognitive function among Chinese adults 65 years and older. Sci Total Environ. 2021;766:142607.

47. Wilker EH, Preis SR, Beiser AS, Wolf PA, Au R, Kloog I, et al. Long-term exposure to fine particulate matter, residential proximity to major roads and measures of brain structure. Stroke. 2015;46:1161–6.

48. Mueller W, Steinle S, Parkka J, Parmes E, Liedes H, Kuijpers E, et al. Urban greenspace and the indoor environment: pathways to health via indoor particulate matter, noise, and road noise annoyance. Environ Res. 2019;180:108850.

49. Brantley HL, Hagler GS, Deshmukh PJ, Baldauf RW. Field assessment of the effects of roadside vegetation on near-road black carbon and particulate matter. Sci Total Environ. 2014;468:120–9.

50. Salmond JA, Williams DE, Laing G, Kingham S, Dirks K, Longley I, et al. The influence of vegetation on the horizontal and vertical distribution of pollutants in a street canyon. Sci total Environ. 2013;443:287–98.

51. Abhijith KV, Kumar P. Field investigations for evaluating green infrastructure effects on air quality in open-road conditions. Atmos Environ. 2019;201:132–47.

52. Baldauf R. Roadside vegetation design characteristics that can improve local, near-road air quality. Transportation Res Part D Transp Environ. 2017;52:354–61.

53. Brugge D, Patton AP, Bob A, Reisner E, Lowe L, Bright OJM, et al. Developing community-level policy and practice to reduce traffic-related air pollution exposure. Environ Justice. 2015;8:95–104.

54. Yang J, McBride J, Zhou J, Sun Z. The urban forest in Beijing and its role in air pollution reduction. Urban For Urban Green. 2005;3:65–78.

55. Selmi W, Weber C, Rivière E, Blond N, Mehdi L, Nowak D. Air pollution removal by trees in public green spaces in Strasbourg city, France. Urban For Urban Green. 2016;17:192–201.

56. Pettit T, Torpy FR, Surawski NC, Fleck R, Irga PJ. Effective reduction of roadside air pollution with botanical biofiltration. J Hazard Mater. 2021;414:125566.

57. Dadvand P, Nieuwenhuijsen MJ, Esnaola M, Forns J, Basagaña X, Alvarez-Pedrerol M, et al. Green spaces and cognitive development in primary schoolchildren. Proc Natl Acad Sci USA. 2015b;112:7937–42.

58. Mason L, Ronconi A, Scrimin S, Pazzaglia F. Short-Term Exposure to Nature and Benefits for Students' Cognitive Performance: a Review. Educ Psychol Rev. 2022;34:609–47. https://doi.org/10.1007/s10648-021-09631-8.

59. Fong KC, Kloog I, Coull BA, Koutrakis P, Laden F, Schwartz JD, et al. Residential greenness and birthweight in the state of Massachusetts, USA. Int J Environ Res Public Health. 2018;15:1248.

60. Gascon M, Triguero-Mas M, Martínez D, Dadvand P, Rojas-Rueda D, Plasència A, et al. Residential green spaces and mortality: a systematic review. Environ Int. 2016;86:60–7.

61. Coleman CJ, Yeager RA, Riggs DW, Coleman NC, Garcia GR, Bhatnagar A, et al. Greenness, air pollution, and mortality risk: a US cohort study of cancer patients and survivors. Environ Int. 2021;157:106797.

62. World Health Organization. Air pollution and child health: prescribing clean air: summary. World Health Organization; 2018. https://iris.who.int/handle/10665/275545.

63. Osborne S, Uche O, Mitsakou C, Exley K, Dimitroulopoulou S. Air quality around schools: Part I-A comprehensive literature review across high-income countries. Environ Res. 2021;196:110817.

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VNM designed the research and drafted the paper. IH, CMK, JEH, MH, JMW provided the data. VNM, JMG, WP, DRG, PK performed the research. All authors contributed to editing the article.

COMPETING INTERESTS

The authors declare no competing interests.

ADDITIONAL INFORMATION

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