

Long-term trends in urban NO₂ concentrations and associated pediatric asthma cases: estimates from global datasets

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Abstract

Background: Levels of nitrogen dioxide (NO₂), a combustion-related air pollutant largely associated with traffic in urban areas, have been changing rapidly due to competing influences of regulation and population and fossil fuel-powered economic expansion. Traffic-related NO₂ is associated with pediatric asthma incidence in epidemiological studies around the world. We aim to assess long-term trends in NO₂ concentrations and NO₂-attributable pediatric asthma incidence in cities globally.

Methods: We estimate global annual average surface NO₂ concentrations at 1km resolution for 1990-2019 by combining land use regression model predictions with NO₂ column densities from the Ozone Monitoring Instrument satellite sensor. We use these concentrations with an epidemiologically-derived concentration-response factor, population, and baseline disease rates to estimate NO₂-attributable pediatric asthma incidence. We explore trends over the last two decades.

Findings: We found diverging regional trends leading to an emerging global convergence in urban NO₂ concentrations globally from 2000-2019. Concentrations are high but declining in high-income countries and low but rising elsewhere. Estimated NO₂-attributable pediatric asthma incidence shows similar trends with decreases of 28-56% in North America, Western and Central Europe, and Australasia, but increases of >50% in Central and South Asia and >100% in Sub-Saharan Africa.

Interpretation: Traffic-related air pollution continues to be an important contributor to pediatric asthma incidence in cities in both developed and developing countries. Divergent experiences of different world regions show that while population growth is worsening NO₂ levels with substantial implications for

children's health in Asia and Africa, rapid and substantial NO₂ declines are possible with effective regulations.

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Research in context:

Evidence before this study: We searched PubMed and Google Scholar databases for studies published in English from the database inception until March 11, 2021, using the search terms ("NO₂" OR "nitrogen dioxide") AND "asthma" AND "trends". Previous studies have reported epidemiological analyses linking changes with asthma to changes in NO₂, or have assessed long-term trends in NO₂ concentrations in some countries or world regions. However, these studies provide little information about how NO₂ concentrations are changing in urban areas all around the world, and the influence those changes have on pediatric asthma incidence. One study published in 2019 showed that over 4 million new pediatric asthma cases, representing ~13% of all pediatric asthma cases worldwide in 2015, could be attributed to NO₂ pollution. Understanding temporal trends in NO₂-attributable pediatric asthma incidence could help inform asthma and air pollution mitigation strategies.

Added value of this study: To our knowledge, this is the first study to have estimated long-term trends in NO₂ concentrations and NO₂-attributable pediatric asthma incidence in urban areas worldwide. We focus on the last two decades, from 2000 to 2019, a period in which NO₂ levels have been changing rapidly around the world due to competing influences of regulation versus population growth and fossil fuel-powered economic expansion. We show that regional trends are diverging, with high but declining concentrations in high-income countries that have regulated NO₂ and low but rising concentrations in other parts of the world where population is expanding. The effect of these contrasting trends is an emerging convergence in urban NO₂ concentrations, and, to a lesser extent, NO₂-attributable asthma incidence, globally.

Implications of all the available evidence: Urban NO₂ concentrations have been trending downward for decades in places that have effective air quality regulations, with benefits for improved children's respiratory health. Even with these improvements, the current levels of NO₂ contribute substantially to pediatric asthma incidence, highlighting that mitigating air pollution should be a critical element of children's public health strategies. For cities that have not benefited from strong local or national air quality management programs, the experience of cities that have such programs demonstrates that addressing combustion-related air pollution can lead to dramatic improvements in air quality and public health over short time frames. Our study also demonstrates the utility of satellite remote sensing for environmental and public health surveillance in urban areas worldwide.

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Data availability: NO₂ concentrations are available at:

https://figshare.com/articles/dataset/Global_surface_no2_concentrations_1990-2020/12968114.

Estimated NO₂-attributable asthma incidence results are available at:

https://github.com/AMohegh/Asthma_no2_urban

Author contributions: SCA, AM, and DG designed the study; PH, AL, and LL provided data; AM carried out the calculations, SCA, AM, DG, MB, PH, AL, and SW contributed input on methods development; SCA and AM wrote the paper; all authors helped interpret results and reviewed the manuscript. AM and DG have verified the underlying data.

Declaration of interests: The authors declare no conflict of interest.

Introduction

Nitrogen dioxide (NO₂), a component of nitrogen oxides (NO_x), is a pervasive air pollutant that is regulated by the U.S. Environmental Protection Agency. NO₂ is a precursor for ground-level ozone and fine particulate matter (PM_{2.5}), other regulated air pollutants that are the leading contributors to air pollution-related mortality.¹ NO₂ is an effective tracer for anthropogenic fuel combustion generally, and traffic specifically.^{2–5} Major anthropogenic NO₂ sources include on-road and non-road transportation tailpipe emissions (including heavy-, medium-, and light-duty vehicles, shipping, and aviation), power plants, industrial manufacturing, and agriculture.^{6–9} In urban and industrialized areas, a lower fraction of NO₂ comes from natural sources like soil and lightning. NO₂ concentration trends can be used to evaluate the efficacy of air pollution regulations, as well as effects of abrupt emission changes (e.g. power plant closures, new oil and gas fields, COVID-19 lockdowns).^{10–13}

In addition to its role in PM_{2.5} and ozone formation, NO₂ itself (often as a marker for the broader traffic-related air pollution mixture) has been linked to several adverse health outcomes including asthma exacerbation.^{14,15} Recent epidemiological studies have also found associations between transportation related air pollutants with new asthma development in children.^{16,17} While the putative agent in the traffic-related air pollution mixture remains unknown, epidemiological studies are relatively consistent in their finding that NO₂ is significantly associated with pediatric asthma incidence, while the evidence for other traffic-related air pollutants (e.g. PM_{2.5}) is more mixed.^{16,17} Previous health impact assessments have linked NO₂ with over four million new pediatric asthma cases each year globally, ~13% of the global pediatric asthma burden.^{18,19} The fraction of new pediatric asthma cases that were attributed to NO₂ ranged substantially higher, up to ~50%, in urban areas.

NO_x emissions and NO₂ concentrations have changed dramatically in response to socioeconomic changes.^{20–24} In the U.S., NO₂ concentrations have dropped ~50% on average from the 1980s to 2010s,²⁵ with even larger drops near major roadways²⁵ and point sources.¹⁰ In the last two decades, U.S. NO_x emissions have been falling ~5% per year as vehicles get more fuel efficient and cleaner and as power plants have been shifting from coal to relatively cleaner fuels (e.g. natural gas).^{11,26,27} NO₂ concentrations have also decreased in Europe, though more slowly.^{28,29} In contrast, NO₂ has been increasing in India,³⁰

the Middle East,³¹ and eastern Europe.²² In China, NO_x emissions peaked around 2011/2012 and subsequently declined.^{32–34}

Previous research on NO₂ temporal trends in urban areas have focused only on a small subset of cities and have not considered its health impacts, precluding the ability to compare urban trends in NO₂ concentrations and associated health burdens in a consistent manner across the world. The global coverage and long continuous record of satellite remote sensing since the 1990s make it possible to track urban NO₂ concentrations globally.^{35–38} Additionally, current satellite instruments operate with substantially higher spatial resolution compared with their predecessors, enabling assessments of intra-urban NO₂ variation and identification of small NO_x emissions sources.^{39,40}

Here we investigate long-term trends of annual average NO₂ concentrations and associated pediatric asthma burdens in urban areas over the past two decades (2000–2019) globally. We first generate a new gridded global surface annual average NO₂ concentration dataset from 1990 to 2019 using satellite data from the Ozone Monitoring Instrument (OMI) to scale NO₂ concentrations estimated from a previously published land use regression (LUR) model⁴¹ to different years. We then use epidemiological concentration-response relationships to estimate the temporal trend in urban NO₂-attributable pediatric asthma cases globally over the last two decades (2000–2019). Finally, we deconstruct the drivers of these trends to explore the influence of NO₂ versus demographic changes.

Methods

We integrated global environmental and demographic datasets available from different sources to generate new estimates of surface NO₂ concentrations and NO₂-attributable pediatric asthma incidence in cities globally. The analysis was done in Python (version 3.6.7).

NO₂ concentrations

We estimated surface annual average NO₂ concentrations at 0.0083 degree resolution (~1 km²) in five-year increments from 1990–2010 and annually from 2010–2019. To begin, we used a previously published NO₂ concentration dataset for the average of 2010–2012, which used land use regression modeling with inputs from road networks and other land use variables, as well as satellite NO₂ column observations from SCIAMACHY and GOME-2.^{4,41} We aggregated NO₂ concentrations from this dataset from its native 100m x 100m resolution globally to 0.0083 degree resolution, which is still a high enough resolution to avoid substantially underestimating NO₂-attributable asthma impacts.¹⁹

Due to the lack of ground measurements in rural areas, the Larkin et al.⁴¹ NO₂ dataset is fine-tuned towards the urban areas and overestimates NO₂ concentrations in rural areas. We therefore applied the Larkin et al.⁴¹ NO₂ concentrations in all 1km x 1km grid cells globally that are categorized as “urban” according to the Global Human Settlement Model grid,⁴² as well as those grid cells that include major roadways⁴¹. For grid cells >5km away from roadways and in rural areas, we developed new NO₂ concentration estimates using NO₂ column observations from the OMI satellite instrument with some adjustments to fill spatial and temporal gaps in the OMI satellite record, and to estimate 24 hour averages from the early afternoon OMI overpass time. The datasets, correction factors, and evaluation against ground monitor observations are described in the Supplemental Material and in Table S1 and Figures S1 through S5.

While the global NO₂ concentration dataset is available for 1990-2019, the estimates for 1990 and 1995 have more uncertainty. Here, we use the results for 2000 to 2019 to explore trends and NO₂-attributable pediatric asthma incidence, as the long-term satellite observational record enhances confidence in results for this period.

NO₂-attributable pediatric asthma incidence

We estimated NO₂-attributable cases of pediatric asthma incidence for 2000-2019 using an epidemiologically-derived concentration response function, following the method used by Achakulwisut et al.¹⁸ and Mohegh et al.¹⁹, as shown in Equation 1:

$$\text{Equation 1: } Burden_{C, AG} = \sum_{\text{Grid cells in } C} Inc_{C, AG} \times Pop_{\text{Grid cell } (i,j), AG} \times (1 - e^{-\beta_{AG} X_{\text{Grid cell } (i,j)}})$$

where C is the country, AG is the age group, Inc is the baseline asthma incidence rate for the age group and country, Pop is the grid-cell population, β is the concentration response factor relating the NO₂ concentration with increased risk of asthma incidence, and X is the grid-cell annual average NO₂ concentration. We perform the health impact calculations at a spatial resolution close to the resolution of the population dataset (1km), since previous work indicates that 1km resolution achieves computational efficiency while still capturing co-location of concentrations and population at the city-scale. Therefore, all other datasets have been re-gridded to match it, including the concentration dataset at ~100m resolution and asthma baseline rates at the country/territory level). We assumed a theoretical minimum risk exposure level (TMREL) of 2 ppb annual average NO₂ concentration, and only estimated NO₂-attributable asthma incidence above that level. We applied a relative risk (RR) of 1.26 per 10ppb increase in annual average NO₂ concentration from Khreis et al.¹⁶

We use population estimates from Worldpop (Tatem 2017) in four different age groups (1-4, 5-9, 10-14, and 15-18 years old) from 2000-2019 at ~1km resolution. National baseline asthma incidence rates are from the Institute for Health Metrics and Evaluation's Global Burden of Disease (GBD) 2019 Study for the same age groups. Boundaries of urban areas are from the GHS-SMOD Urban Centre dataset which provides shapefiles of boundaries of over 13,000 urban clusters globally for the year 2015 (latest year).⁴² We use the 2015 urban boundaries for all years of our analysis. Here, we consider grid-cells to be part of an "urban cluster" if they are included in "urban" and "sub-urban" areas in the GHS-SMOD dataset, which narrows the selection to areas with more than 300 people per km² that are part of clusters with more than 5000 population. We consider all other gridcells to be "rural". Regional definitions are consistent with the Global Burden of Disease 2019 Study.¹

Drivers of change

To disentangle the contribution of exposure, population, and disease rates to NO₂-attributable pediatric asthma cases, we performed four sets of simulations: a control scenario, where we calculate the estimated asthma cases for each year, and three sets of scenarios in which we revert one contributing parameter back to the base year of 2000. Using a combination of these simulations, we estimate the contribution of each parameter in the overall changes in estimated asthma cases between 2000 and

each year of analysis through 2019. The contribution of each parameter to that change in NO₂-attributable asthma cases between years is calculated using the formulas presented in the supplemental document. This method allows for calculating ratios for each parameter that add up to the scaling factor between the year 2019 and 2000, and due to its logarithmic nature, it is consistent with the original health impact function, which is a multiplication of three parameters. This approach is described in more detail in the Supplemental Material.

Role of the funding source

The funders of the study had no role in study design, data collection, data analysis, data interpretation, or writing of the report. SA and AM had full access to all the data in the study and had final responsibility for the decision to submit for publication.

Results

When averaged globally, urban annual average NO₂ concentrations decreased by 13% between 2000 and 2019 (Figure 1). We estimate that in 2000, the global urban population weighted average concentration was 12.2 ppb, it increased between 2006 and 2011, and then slowly declined from 2012 to a level of 10.6 ppb in 2019. However, these global trends mask steeper, and contrasting trends in different world regions. NO₂ concentrations in high-income cities were the highest among all regions in 2000, and remained above the global average throughout the time period, even after the steady, nearly monotonic decline of 35% in these cities from 2000 to 2019. In contrast, concentrations have risen in South Asia by 16% and in Sub-Saharan Africa by 11% over this period, though concentrations in these regions remain the lowest of cities in all other world regions in 2019. Cities in other world regions experienced only modest changes in NO₂ concentrations from 2000 to 2019. Even these large regional groupings of cities obscure contrasting temporal trends between sub-regions. For example, within the high-income region, concentrations dropped by nearly 53% in North America, but only by 35% and 39% in Western Europe and Australasia, respectively. In high-income Asia Pacific cities, concentrations rose dramatically in the 2000s and then declined steadily from 2005 to 2019, and are now at approximately the same level as in 2000. Another region with contrasting sub-regional trends is Central and Eastern Europe and Central Asia, where concentrations have dropped by a large amount in Central and Eastern Europe but have risen steadily in Central Asia. Similar trends can be seen when including all urban and rural areas together (Figure S7).

We estimate that 1.85 million new pediatric asthma cases were attributed to NO₂ pollution globally in 2019 (in both urban and rural areas, Figure S8), approximately 65% of which occurred in urban areas (Figure S9). Similar to concentrations, estimated NO₂-attributable pediatric asthma cases changed very little from 2000 to 2019 (Figure 2). Regional trends in estimated NO₂-attributable asthma incidence largely follow the NO₂ concentration trends, though with less cross-over in rankings between the world regions over time. For example, the high-income region had the most NO₂-attributable asthma incidence in 2000, and it still had the most in 2019, despite declining trends. Similarly, Central and Eastern Europe and Central Asia, South Asia, and Sub-Saharan Africa had the least NO₂-attributable asthma cases throughout 2000-2019, despite increasing trends over time. Sub-regional trends in NO₂-attributable

asthma incidence show a similar pattern as sub-regional trends in NO₂ concentrations. Considering individual countries, the 2019 to 2000 ratio of urban NO₂-attributable asthma cases is relatively consistently positive throughout India, generally negative with increases in some areas within the United States and Western European countries, and very mixed within China, South American countries, and South Africa (Figure 3). While approximately two-thirds of NO₂-attributable asthma cases occurred in urban areas, there is wide variation across regions driven largely by differences in urban vs rural population in these regions (Figure S4).

The estimated trends in urban NO₂-attributable pediatric asthma incidence are driven by simultaneous and often competing changes in NO₂ concentrations, pediatric population size, and asthma incidence rates over time (Figure 4). We find that these drivers of the NO₂-attributable asthma trends are inconsistent globally, though some world regions show similar influences. No region that has experienced a decline in estimated NO₂-attributable asthma cases has done so without a drop in NO₂ concentrations. Similarly, all regions that experienced an increase in NO₂ concentrations saw an increase in estimated NO₂-attributable asthma cases. Australasia and high-income Asia Pacific are the only regions where concentrations, population, and disease rates all contribute to the overall decline in NO₂-attributable asthma cases. The opposite occurred in Central, South, and Southeast Asia, as well as in North Africa and the Middle East, where concentrations, population, and disease rates all contributed to an increasing trend in NO₂-attributable asthma cases. High-income North America, Western Europe, and several other regions experienced competing effects of reduced NO₂ concentrations but increases in asthma incidence rates and/or pediatric population size. These competing influences changed over time, such that declining concentrations have a larger influence over time in North America and Southern Latin America, while population growth becomes more of a factor in North Africa and Middle East and Southern Sub-Saharan Africa (Figure S10). Changing baseline pediatric asthma rates also influence the trends in some regions, in particular East Asia and Southern Sub-Saharan Africa.

Discussion

We found an emerging convergence in NO₂ concentrations globally, with high but declining concentrations in high-income countries that have regulated NO₂ and low but rising concentrations in other parts of the world where population is expanding. Estimated urban NO₂-attributable pediatric asthma incidence also appears to be beginning to converge globally, though with wider separation between world regions compared with regional average concentrations, driven by differences in population and asthma rates between countries. In no region did NO₂-attributable asthma cases decline without declining NO₂ concentrations, while population growth almost always contributed to rising trends in NO₂ asthma impacts. The influence of baseline pediatric asthma incidence rates was inconsistent between regions.

Our estimated NO₂ concentration trends are consistent with recent studies that used satellite data to investigate global NO₂ concentration trends during 2004-2018 for the U.S., Europe, China, India, and Japan^{21,22,43}. Qu et al.⁴⁴ also showed a similar decrease in U.S. NO₂ concentrations based on ground observations, satellite data, and modeling outputs from 2006-2016, and Henneman et al.²⁵ showed a similar decrease using ground observations from 1980-2020.

Our estimate of 1.84 million NO₂-attributable new pediatric asthma cases globally (in both urban and rural areas) in 2019 is less than half of the 4.2 million found by Achakulwisut et al.¹⁸ for 2015. Our estimates for India and the U.S. are 35% lower and 53% higher than our previous estimates, respectively¹⁹. Several factors explain this discrepancy. First, our new NO₂ concentration dataset corrects for a high NO₂ concentration bias in rural areas. This change results in lower contributions of NO₂ to asthma incidence in rural areas, particularly in countries with larger rural populations like India. Second, compared with previous GBD versions, GBD 2019 baseline asthma rates are much lower, except for high income areas such as the U.S. For example, the global baseline asthma incidence rate for people <20 years old in 2015 (the year analyzed by Moheghe et al.¹⁹) decreased from 1055 to 828 cases per 100,000 in the GBD 2017 (GBD 2017) vs. GBD 2019 Study (GBD 2019). This difference is even larger in China and India, where pediatric asthma incidence rates in the GBD 2019 Study were ~72% of their value in the GBD 2017 Study. U.S. baseline pediatric asthma incidence rates increased from 1257 to 2750 cases per 100,000 in the GBD 2017 vs. GBD 2019 Study. Changes in baseline asthma rates have approximately proportional effects on estimated NO₂-attributable pediatric asthma cases. Finally, as we have shown, NO₂ concentrations are changing rapidly in different directions depending on the region and city, with the same directional effect on estimated NO₂-attributable pediatric asthma cases.

While we sought to create a new NO₂ dataset that leverages the advantages of different data sources, our new NO₂ concentrations are still uncertain. First, many cities are still lacking ground NO₂ monitors, limiting our ability to calibrate and evaluate concentrations in these regions.⁴¹ Urban NO₂ concentrations therefore have more certainty in North America, Europe, and Asia, compared with Africa and South America. Second, rural NO₂ concentrations are uncertain in all regions globally due to the disproportionate location of ground monitors in urban areas. Third, our approach for scaling NO₂ concentrations from the 2010-2012 average LUR model to other years assumes that the land use predictors of that regression model are unchanged over time. This assumption is likely supported by slow changes in road density and volume and urban form from one year to the next, but over the two decades time frame explored here, some land use evolution is likely. The directional impact of these uncertainties and limitations on results is currently unknown.

Estimated NO₂-attributable asthma incidence is heavily influenced by baseline disease rates, which vary considerably within individual countries.⁴⁵ While we used national pediatric asthma rates, asthma prevalence differs between urban and rural areas. Living in urban areas has been associated with increased risk of asthma prevalence in low- and middle-income countries⁴⁶ and asthma-related emergency department visits and hospitalizations in the U.S.⁴⁷ Our method also does not capture temporal trends in asthma rates in cities specifically, which may differ from the broader national rate changes. If asthma prevalence tends to be higher in urban areas compared with national averages, we may have underestimated NO₂-attributable asthma cases.

Despite these uncertainties and limitations, our results demonstrate the important influence of combustion-related air pollution on children's health in cities globally. In places that have effective air quality management programs, NO₂ concentrations have been trending downward for decades, with benefits for improved children's respiratory health. Even with these improvements, the current levels of NO₂ contribute substantially to pediatric asthma incidence, highlighting that mitigating air pollution should be a critical element of children's public health strategies. For cities that have not benefited from strong local or national-scale air quality management programs, the experience of cities that have such programs demonstrates that addressing combustion-related air pollution can lead to dramatic

improvements in air quality and public health over short time frames. These air quality improvements can be achieved through either end-of-pipe emission control technologies or avoiding the combustion in the first place, which would have additional benefits from reduced greenhouse gas emissions.

Our study also highlights the value of satellite remote sensing and statistical models for tracking NO₂ pollution and for environmental health surveillance at local, national, and global scales. The combination of methods offers strengths beyond what is possible from each technique alone: a long and consistent observational record of NO₂ column densities from satellite remote sensing with the high spatial resolution of the surface concentration predictions from the LUR model. Future studies may seek to leverage these data sources and others, including mobile monitoring, distributed networks of ground sensors, and chemical transport models, to further improve upon the accuracy and spatiotemporal resolution of NO₂ concentration estimates. Further, our study shows the importance of considering not just concentrations, but how demographics change over time, for understanding air pollution health risks. Improved and more widely accessible information about disease rates, and capturing population distribution and movement, will enable more accurate and highly resolved air pollution health impact assessments.

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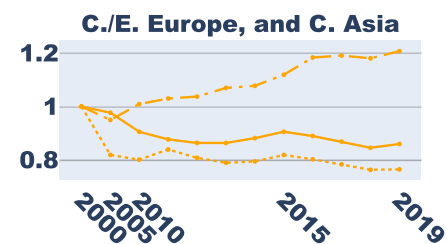
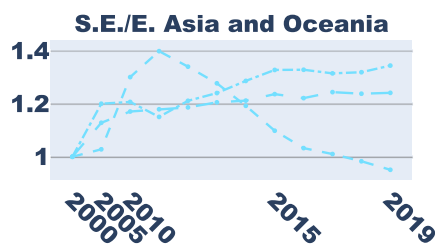
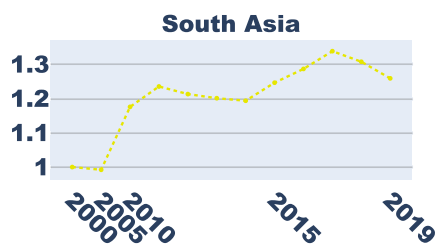
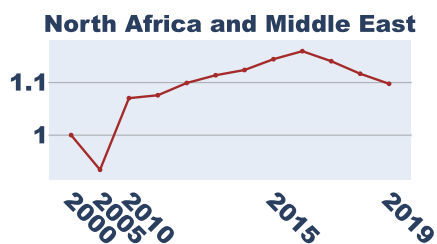
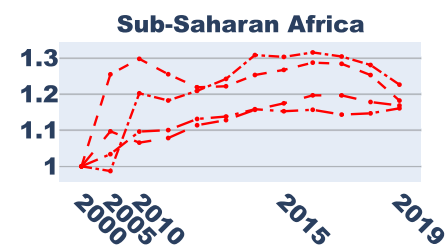
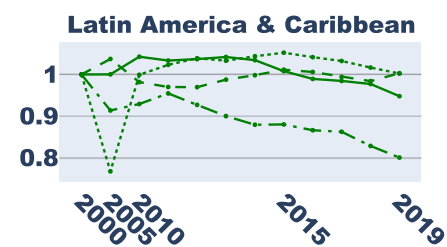
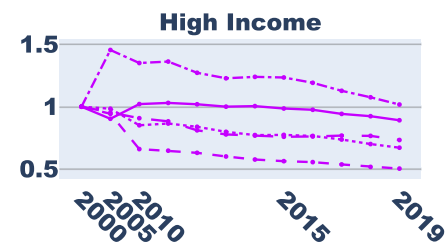
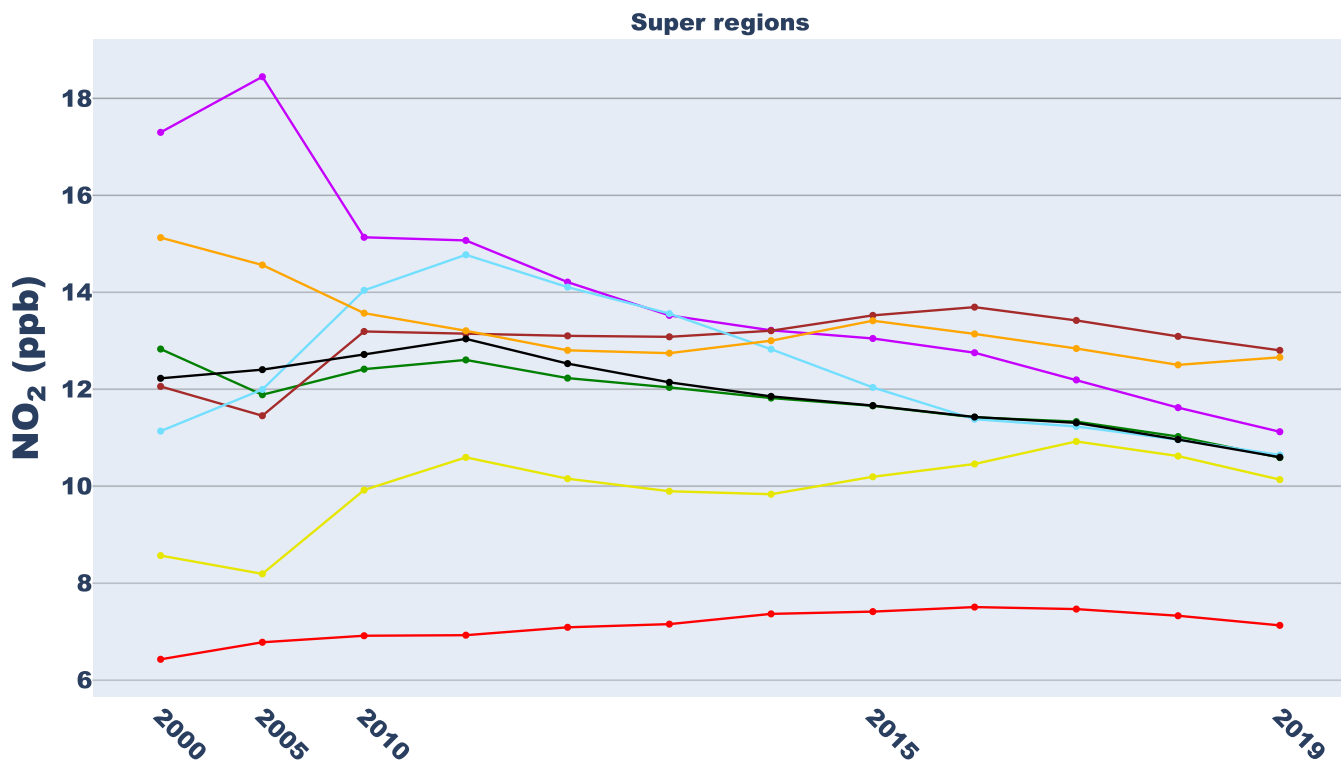
Figure captions

Figure 1. Trends in annual average NO₂ surface concentrations in urban areas in various world regions. The figure presents regional average concentrations aggregated for GBD super regions in form of NO₂ concentrations (main panel) and aggregated results for GBD regions in each super region in form of relative changes from the year 2000 (side panels). Regional definitions are shown in Figure S5. China is included in East Asia and India is included in South Asia. H.I. = High Income; L.A.C. = Latin America & Caribbean; S.S.A. = Sub-Saharan Africa; N.A.,M.E. = North Africa and Middle East; S.A. = South Asia; SE.A., E.A., O. = Southeast Asia, East Asia, and Oceania; C.E., E.E., C.A. = Central Europe, Eastern Europe, and Central Asia.

Figure 2. Trends in estimated urban NO₂-attributable pediatric asthma incidence in each super-region (absolute magnitude, main panel) and sub-region (relative changes from 2000, side panels). The legends are the same as Figure 1.

Figure 3. Relative changes in estimated NO₂-attributable pediatric asthma incidence between 2000 and 2019 globally (including both urban and rural areas). Lack of color indicates no NO₂-attributable pediatric asthma incidence as NO₂ concentrations were <2 ppb, the low concentration threshold for estimating attributable asthma incidence.

Figure 4. Contribution of population, baseline asthma rates, and NO₂ concentrations to changes in estimated NO₂-attributable pediatric asthma incidence between the year 2000 and 2019 for each sub-region. Since meteorological factors can influence comparisons of concentration contributions when comparing two years, we show contributions by year for several regions in Figure S3. Percentage labels indicate the net change.



— Global — H.I., Southern Latin America - - - H.I., Western Europe - - - H.I., High-income North America — H.I., Australasia
 - - - H.I., High-income Asia Pacific — L.A.C., Central Latin America — L.A.C., Tropical Latin America
 - - - L.A.C., Andean Latin America - - - L.A.C., Caribbean - - - S.S.A., Southern Sub-Saharan Africa
 — S.S.A., Western Sub-Saharan Africa - - - S.S.A., Central Sub-Saharan Africa - - - S.S.A., Eastern Sub-Saharan Africa
 — N.A., M.E., North Africa and Middle East - - - S.A., South Asia — SE.A., E.A., O., East Asia — SE.A., E.A., O., Southeast Asia
 — SE.A., E.A., O., Oceania — C.E., E.E., C.A., Central Asia — C.E., E.E., C.A., Eastern Europe - - - C.E., E.E., C.A., Central Europe

