

Lockdown caused by COVID-19 pandemic reduces air pollution in cities worldwide

Asheshwor Man Shrestha¹, Uttam Babu Shrestha^{1,*}, Roshan Sharma², Suraj Bhattarai¹, Hanh Ngoc Thi Tran³, Maheswar Rupakheti⁴

¹Global Institute for Interdisciplinary Studies, Kathmandu, Nepal

²Centre for Urban Research, RMIT University, Melbourne, VIC, Australia

³School of Marketing and Management, The University of Adelaide, Adelaide, SA, Australia

⁴Institute for Advanced Sustainability Studies, Potsdam, Germany

*Corresponding author: Uttam Babu Shrestha (Email: ubshrestha@yahoo.com)

Abstract

Drastic measures such as lockdown taken by countries worldwide to contain spread of COVID-19 have influenced air pollution dynamics substantially, at a planetary scale. Several media reported the lockdown induced air pollution reduction based on quickly assembled satellite observations. However, a global-scale analysis of such reduction in air pollution backed by quality data collected from ground stations across the world and the effects of lockdown to it are missing. Here, we investigate changes in concentrations of six air pollutants: PM_{2.5}, PM₁₀, O₃, SO₂, CO, and NO₂ in 40 cities between February, March 2019 and 2020. The mean monthly concentrations of PM_{2.5} and PM₁₀ in February and March of 2020 were found consistently lower than in 2019 in most of the cities. After lockdown, declines of anthropogenic pollutants such as NO₂, CO, PM_{2.5} and PM₁₀ concentrations were seen in 19, 9, 8 and 7 cities respectively. Improvement in air quality following lockdown was observed in the world's most polluted cities including Bangalore, Beijing, Bangkok, Delhi, and Nanjing, as well as the world's major trade centers including New York, London, Paris, Seoul, Sydney, and Tokyo. More visible reduction of air pollution remains to be seen if the lockdown is prolonged. Nevertheless, such reductions are expected to be temporary because the levels are likely to go up again once the situation gets back to normal.

1. Introduction

The COVID-19 pandemic has impacted every aspect of human life and the global economy. As of April 15, 2020, the coronavirus (COVID-19) has infected over 1.92 million people and caused ~127,000 deaths in 185 countries and territories across the world (<https://coronavirus.jhu.edu/map.html>). The number of new cases and deaths is increasing at an alarming rate with no signs of control yet, making the estimates of its economic and other impacts uncertain. The global cost of the COVID-19 pandemic could range from \$2.0 to 4.1 trillion, equivalent to a loss of between 2.3% and 4.8% of global gross domestic product (ADB 2020). It will make more than 25 million people unemployed by the end of 2020 depending on how quickly the economy recovers in the second half of the year and how effectively policy measures will boost labour demand (ILO 2020).

Depending on the level of COVID-19 impact in each country, as well as country-specific situations and capacity, the world's governments are adopting different levels of interventions including travel restrictions and lockdown to contain the spread of the highly contagious virus. The lockdown restricting movement of people and shutting down business and industrial establishments was first imposed on January 23, 2020 in the city of Wuhan in China where the first case of this deadly virus was reported and later followed by other countries (Jing 2020). From March 25, 2020, India, the second-most populous country in the world with 1.3 billion people, went into lockdown (Chandrashekhara 2020). Although lockdown ended in Wuhan, China on April 8 after 76 days (Kuo 2020), more than 3.9 billion people or a half of the world's population living in 90 different countries around the world are currently under containment as a measure to maintain social distancing (Sandford 2020). The extent of lockdown varies across countries and cities around the world. Nonetheless, the lockdown has halted quite significantly all forms of transport (flights, trains, automobiles), factories, shops, markets, and other economic and social activities.

Despite the severe impacts of lockdown on people's social life, global mobility and economy, there are reports that lockdown has temporarily improved environmental conditions. For example, the lockdown improved the air quality of the 103 cities in India — the most polluted

country in the world which has 21 out of the 30 world's most polluted cities (Singh and Chakraborty 2020). Air pollution is the single largest environmental health risk in the world; ca. seven million people die prematurely every year globally as a result of exposure to air pollution, more than 91% of the world's population lives in places where air quality exceeds the World Health Organization's guidelines (WHO 2020). PM_{2.5} is a common cause for adverse health outcomes such as chronic obstructive pulmonary disease (COPD) and lower respiratory infection (LRI) causing death of nearly three million people globally (GBD 2018). NO₂ is the leading source of childhood asthma in urban areas globally (Achakulwisut et al. 2019). Likewise, acute to chronic exposures to SO₂, CO, and O₃ can harm the human respiratory system, causing breathing problems including occasional death (Chen et al. 2007, Brown 2009). Air pollution also causes adverse impacts on society, economy, and the environment including climate change. It is in fact a major public health, environmental and developmental challenge of our time (Haines et al. 2017). Therefore, understanding this temporary improvement in air quality at the planetary scale provides a unique opportunity to study processes and implications of policy changes to reduce air pollution in future.

There are media reports, primarily based on satellite images, about reduction of air pollution due to the lockdown globally, and some recent scientific studies conducted in a small number of countries (He et al. 2020) and cities (Cadotte 2020) indicated such reductions. While satellite images are useful to monitor air pollution, cautions should be taken to prevent the erroneous interpretation of satellite images (see <https://atmosphere.copernicus.eu/flawed-estimates-effects-lockdown-measures-air-quality-derived-satellite-observations>). To the best of our knowledge, there is no global-scale examination of the impact of COVID-19 induced lockdown on air quality. Here, we provide the first global insight on air pollution by utilizing vast networks of ground-based air quality monitoring stations from 40 major cities from 35 countries across the six continents. The air quality data is collected from 301 ground-based stations for February and March of 2019 and 2020 to compare both monthly and yearly changes post-COVID lockdown. We also examine the effect of lockdown on the daily concentrations of air pollutants of the last 100 days from January 1 to April 9, 2020.

2. Materials and Methods

2.1. Selection of the world's cities

We selected 40 major cities around the world (Figure 1) based on (i) the historical level of air pollution in the city (a mix of cities with dirty and clean air), (ii) reported COVID-19 cases in that city (a mix of the high and low impacted city), (iii) geographical setting and developmental stage (balance between high, middle, and low income countries of all continent), (iv) city lockdown history (early and late beginnings of lockdowns), and (v) availability of air pollution data. We used the World Air Quality Index portal (WAQI) (www.aqicn.org) to download daily records from 301 air pollution monitoring stations spread across 40 selected cities of 35 countries. We deliberately exclude cities around the equator where air quality is often influenced by the emissions from forest fires, and cities near the deserts that often get hit by sand and dust storms. Therefore, most of the cities selected for this study are affected heavily by anthropogenic emissions, and a reduction in air pollution is expected in those cities as a result of lockdown.



Figure 1. Selected cities used for analysis of changes in the concentrations of air pollutants.

2.2. Selection of air pollutants

Normally, air quality is evaluated by measuring the atmospheric concentrations of six pollutants: fine particulate matter (PM_{2.5}, mass concentration of particles with diameters ≤ 2.5 μm), coarse particulate matter (PM₁₀, mass concentration of particles with diameters ≤ 10 μm), ground-level ozone (O₃), nitrogen dioxide (NO₂), sulfur dioxide (SO₂), and carbon monoxide (CO) (EPA 2014). All of them except O₃ have primary sources, such combustion sources as the automobiles and industries, while O₃ is formed in the atmospheric from precursor gaseous species, emitted mostly from primary sources, in presence of sunlight. Therefore, we selected PM_{2.5}, PM₁₀, NO₂, O₃, SO₂ and CO to investigate the changes in air quality among cities before and during lockdowns.

2.3. Data and analysis

We used publicly available air quality data provided by the World's Air Pollution: Real-time Air Quality Index (WAQI project) which curates air quality data collected from >12,000 ground-based air quality monitoring stations (primarily located at/near the US embassies and consulates) situated in 1,000 major cities in more than 100 countries in six continents (<https://aqicn.org/>). It must be noted that the data from WAQI are from the stations managed by government agency or any institution of the respective country and may not be always fully validated. We used daily concentrations of air pollutants collected from multiple WAQI stations where available for further analysis (Table 1). We collected the lockdown dates for the selected cities from local and global media reports (Table 1). By 'lockdown', we meant restrictions on vehicles, commercial flights except for cargo and charter, and prohibitions of commercial activities, except for essential services, resulting in a temporary shutdown of air pollution sources from these activities. Although we cross-verified the accuracy of the starting dates of lockdown by looking at multiple media reports, there may be some discrepancies due to any errors in media reporting and differences in time zones. We assume air quality in February and March of 2019 was under the business as usual conditions and hence considered normal and change in air quality in February and March of 2020 is caused primarily due to the country-specific response to containing the spread of COVID-19. Therefore, we compared the mean monthly concentration of the six air pollutants of February and March of 2019 with February and March of 2020 using

welch two sample t-test. As dates of lockdown vary, we also examined the daily changes in concentrations of these air pollutants from January 1 to April 9 (100 days) to investigate the effect of lockdown on their daily concentrations of air pollutants through linear regression using lockdown as dummy variable.

3. Results and discussions

A consistent reduction in PM_{2.5} concentration was observed among 17 cities in February 2020 compared to February 2019, and 14 cities (increase in Oslo, Vienna, Warsaw) in March 2020 than that of March 2019 (Figure 2). Maximum decline in mean monthly concentration of PM_{2.5} in February was observed in Vienna (-57.1%), Paris (-53.2%), Amsterdam (-47.5%), London (-45.6%), and Dublin (-45.2%). It is worth mentioning here that in absolute term these cities already have cleaner air, i.e., PM_{2.5} concentrations are very low compared to heavily polluted cities. Likewise, in March, a maximum decline was seen in New York (-37.0%), Sydney (-35.1%), Abu Dhabi (-32.4%), Nanjing (-31.3%), Ulaanbaatar (-25.6%). In the cities, where anthropogenic activities are a major source of emissions of air pollutants sharp declines in the PM_{2.5} levels were seen, notably in highly polluted cities like Bangalore, Bangkok, Beijing, Delhi, Nanjing, Wuhan, Ulaanbaatar in Asia and also in developed cities like Madrid, New York, Paris, Seoul, Sydney, Tokyo. In contrast, an increase in PM_{2.5} concentrations was observed in Addis Ababa (lockdown on April 1 2020) and Ho Chi Minh City (lockdown on March 31, 2020). An increase in February PM_{2.5} concentration and decrease in March in Bangkok and Kathmandu are likely due to the lockdown that started in the last week of March. In regards to daily PM_{2.5} concentrations from January 1, 2020 to April 9, 2020, eight cities Amsterdam, Bangalore, Delhi, Lima, Nanjing, New York, Ulaanbaatar, Wuhan showed significant decrease (linear regression, $p < 0.05$) while there was an increase in Buenos Aires, Santiago and Vienna after accounting the lockdown imposed date (Figure 3).

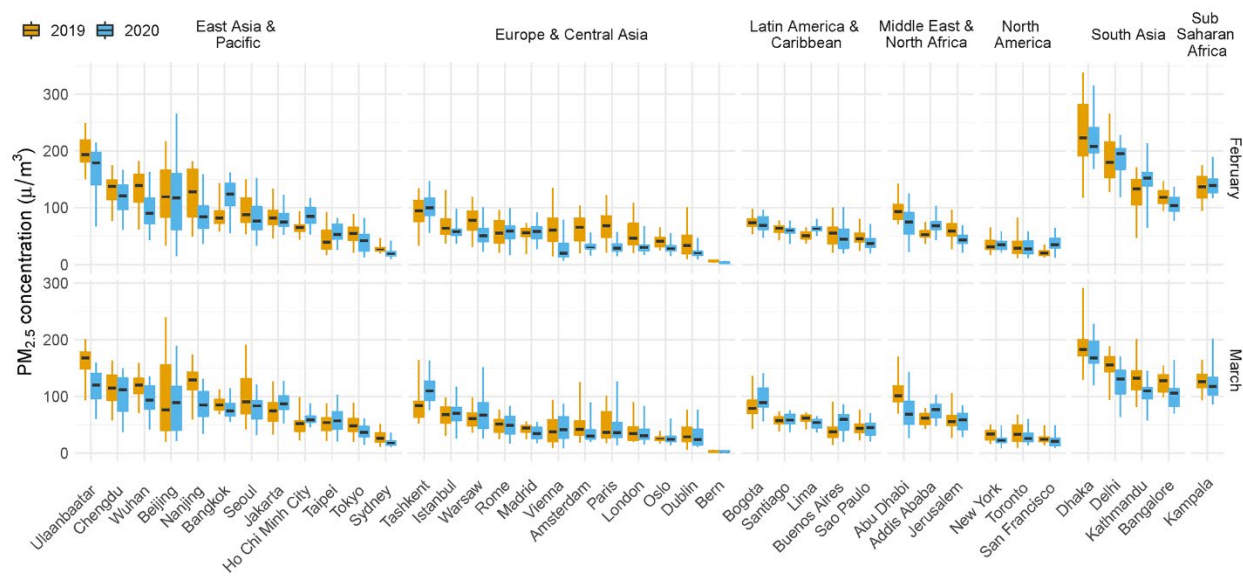


Figure 2. The box and whisker plots of daily average PM_{2.5} concentrations during February-March 2020 and 2019. The top and bottom of each box represents 75th percentile and 25th percentile respectively and the upper and lower whisker represents 90th percentile and 10th percentile respectively. The horizontal bar in each box represents the median.

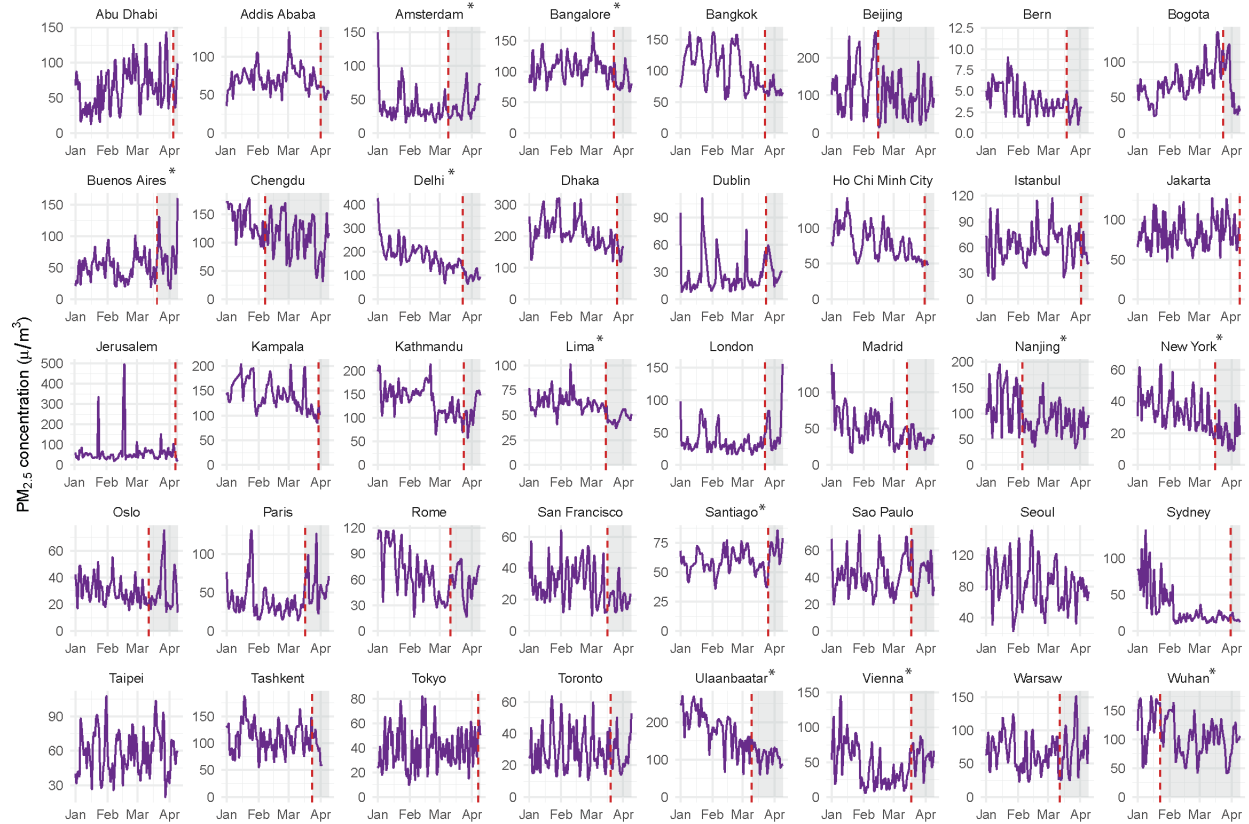


Figure 3. Time series of daily PM_{2.5} concentrations from January 1 to April 9, 2020 in 40 cities across the globe. Vertical red dashed line shows the lockdown date and * denotes significant linear trends.

Only 29 cities have complete PM₁₀ concentrations data of which 16 cities showed a statistically significant decline in PM₁₀ in February 2020 compared to February 2019 and 11 cities showed similar declines in March 2020 than in March 2019. Maximum decline in mean monthly concentration of PM₁₀ in February was observed in Vienna (-60.7%), Bern (-53.3%), Paris (-52.7%), Warsaw (-45.9%) and Dublin (-44.3%). Maximum declines were seen in Ulaanbaatar (-43.4%), Delhi (-32.5%), Nanjing (-31.8%), Wuhan (-26.8%), and Lima (-25.7%) in March. A consistent decline in PM₁₀ in both February and March of 2020 was observed in 7 cities: Bangkok, Beijing, Lima, Nanjing, Tokyo, Ulaanbaatar and Wuhan (Figure 4). The time series of daily PM₁₀ concentrations of the last 100 days in 2020 showed that after the lockdown, a statistically significant decrease in PM₁₀ concentrations was observed in seven cities namely Bangalore, Beijing, Delhi, Lima, Nanjing, Ulaanbaatar, and Wuhan while Kathmandu and

Vienna both saw increases (Figure 5). The local meteorology and seasonal forest fire might have played a role in such increase, particularly stronger wind speeds lifting soil and road dust into air in Kathmandu and stronger winds diluting PM₁₀ concentration in February in Vienna.

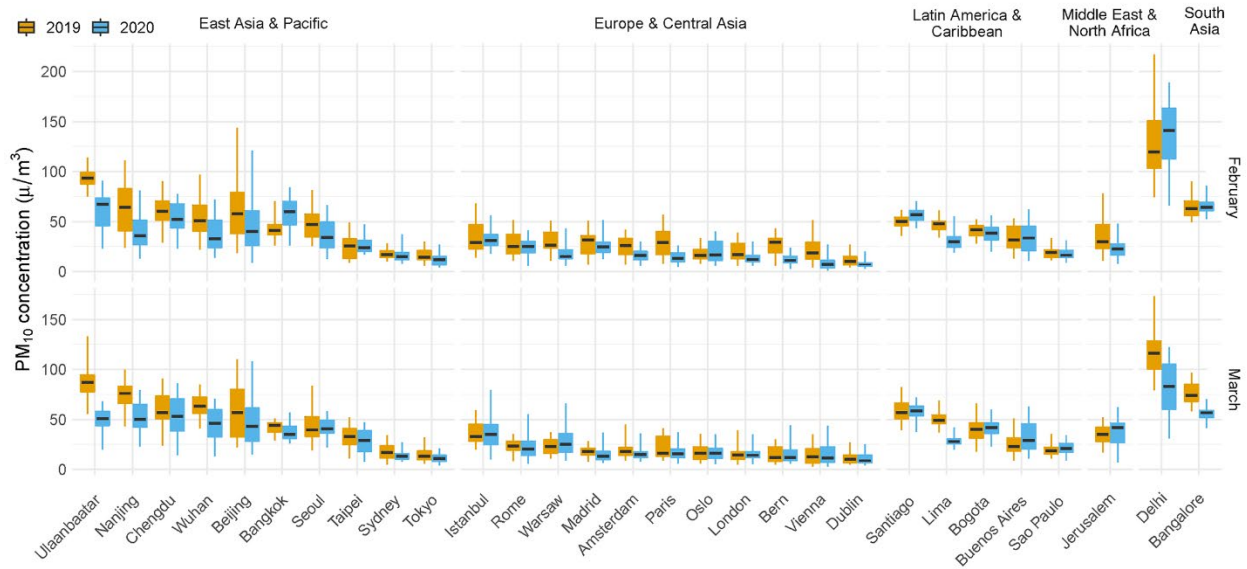


Figure 4. The box and whisker plots of daily average PM₁₀ concentrations during February-March 2020 and 2019. The top and bottom of each box represents 75th percentile and 25th percentile respectively and the upper and lower whisker represents 90th percentile and 10th percentile respectively. The horizontal bar in each box represents the median.

Out of the 31 cities with complete data, decreased NO₂ concentration in February and March 2020 was observed in 10 and 17 cities respectively (Figure 6). There are mixed trends (increases in some cities while decreases in others) in NO₂ for both months; an increase in four cities Wuhan (+62.0% in February, +11.3% in March), Nanjing (+35.1% in February, +62.4% in March), Chengdu (+30.9% in February, +73.9% in March), Warsaw (+4.0% in February, +8.2% in March) while a decrease in five cities; Madrid (-33.3% in February, -47.5% in March, Paris (-

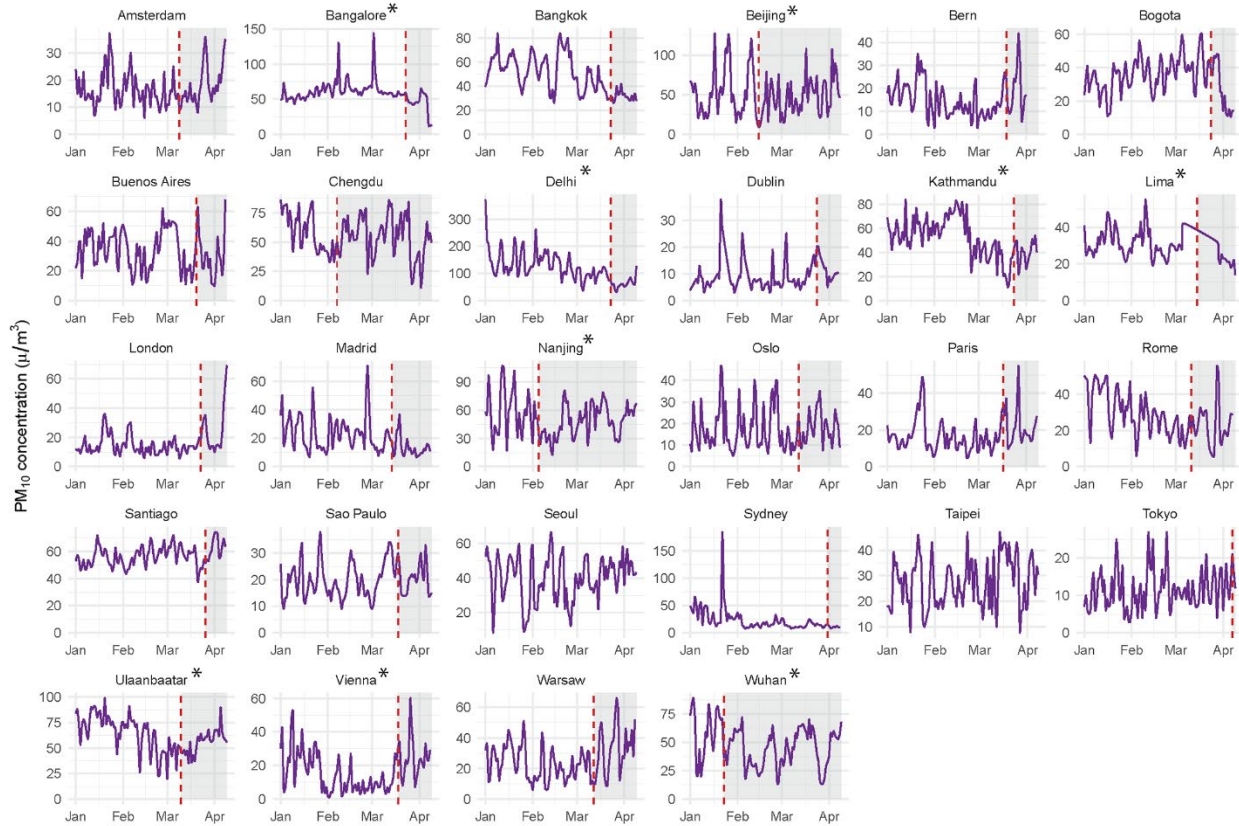


Figure 5. Time series of daily PM₁₀ concentrations from January 1 to April 9, 2020 in 28 cities across the globe. Vertical red dashed line shows the lockdown date and * denotes significant linear trends.

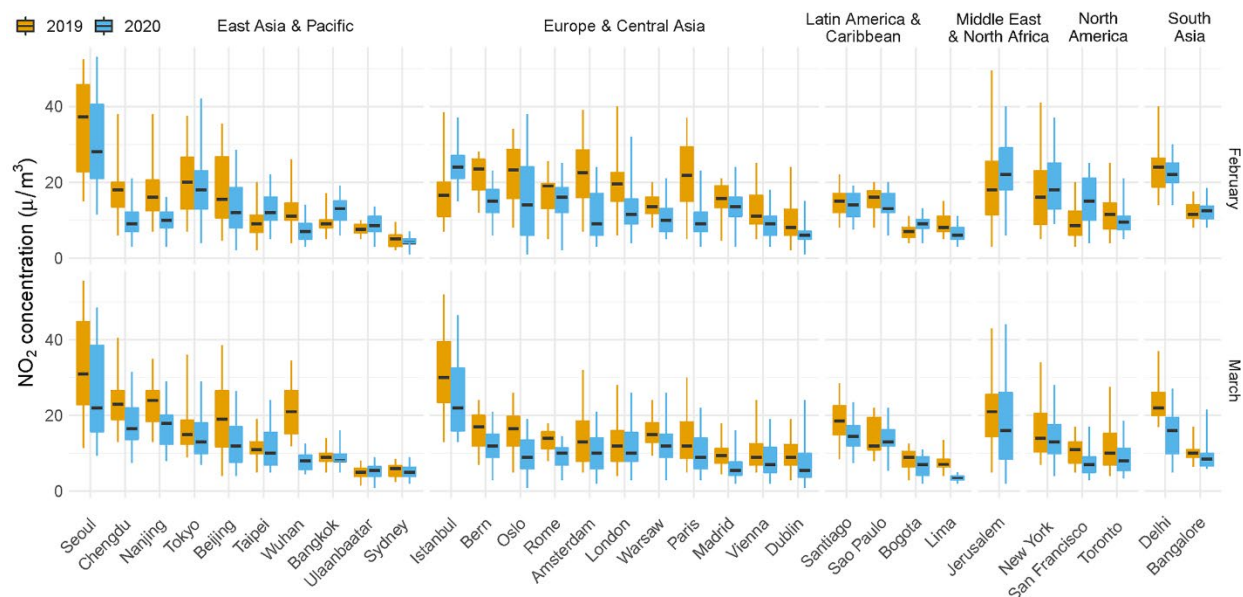


Figure 6. The box and whisker plots of daily average NO₂ concentrations during February-March 2020 and 2019. The top and bottom of each box represents 75th percentile and 25th percentile respectively and the upper and lower whisker represents 90th percentile and 10th percentile respectively. The horizontal bar in each box represents the median.

33.1% in February, -5.6% in March), Oslo (-28.3% in February, -39.7% in March), Bern (-27.0% in February, -19.1% in March), Vienna (-18.1% in February, -4.4% in March), Lima (-4.6% in February, -26.7% in March). However, the significant decline (linear regression, $p < 0.05$) in NO₂ was seen in the daily NO₂ concentration records after the lockdown was imposed; a significant decline in NO₂ concentration was observed in 19 cities (Amsterdam, Bangalore, Beijing, Bern, Bogota, Chengdu, Delhi, Lima, Madrid, Nanjing, New York, Oslo, Paris, Rome, San Francisco, Santiago, Sao Paulo, Warsaw, Wuhan) (Figure 7). Three Chinese cities Chengdu, Nanjing and Wuhan have witnessed consistent increase in NO₂ concentrations in Feb-Mar of 2020 than that of Feb-Mar of 2019 but the daily NO₂ concentrations in these cities were also significantly reduced after lockdown.

Comparison of O₃ concentrations between February 2020 vs. February 2019 and March 2020 vs. March 2019 from available data of 26 cities showed a mixed trend (Figure 8). O₃ concentrations increased in 11 cities in February and in 5 cities in March. While it decreased in 3 cities in February and 4 cities in March. Lima (-42.5%) has maximum decrease followed by Bangkok (-17.6%) in February and Bangkok (-39.5%) and Bangalore (-10.6%) in March. On the other hand, Tokyo (+31.9% in February, +15.2% in March), Madrid (+26.9% in February, 37.8% in March),

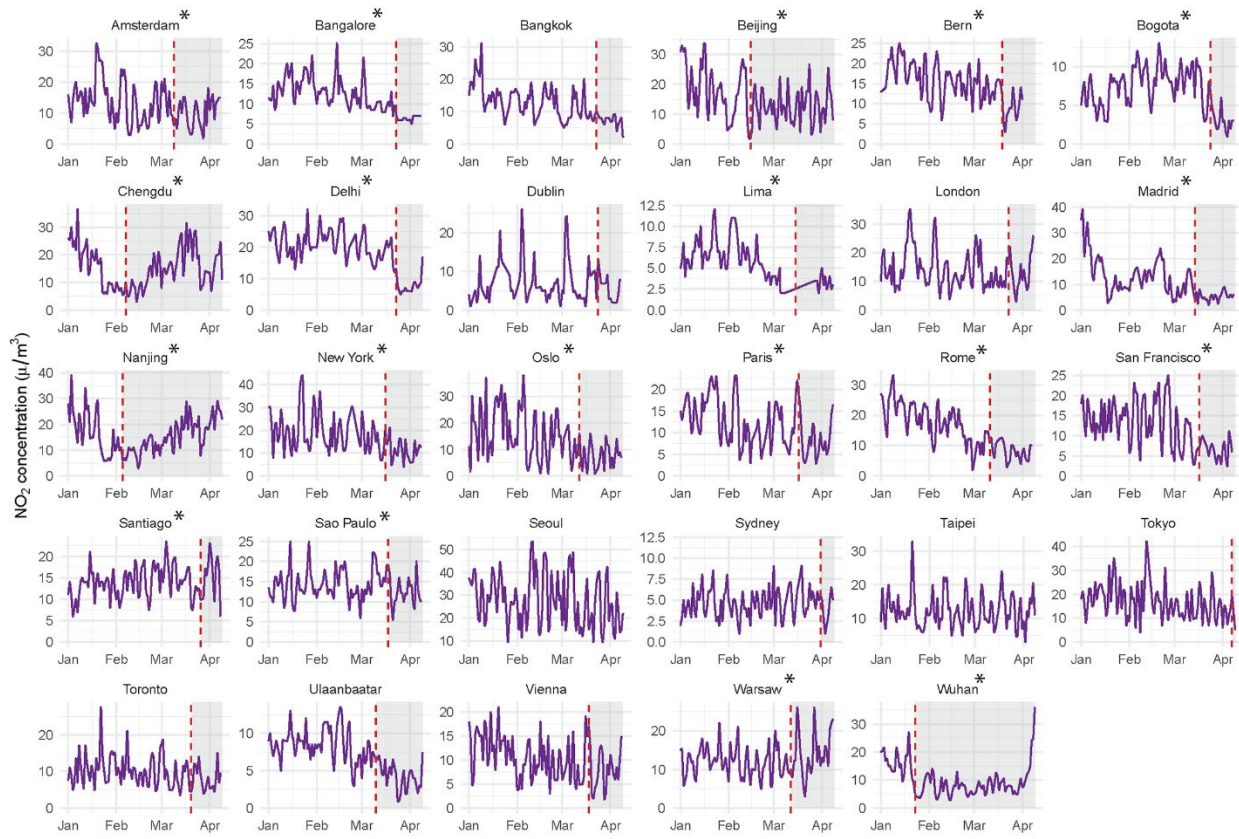


Figure 7. Time series of daily NO_2 concentrations from January 1 to April 9, 2020 in 29 cities across the globe. Vertical red dashed line shows the lockdown date and * denotes significant linear trends.

and Toronto (+17.8% in February, +3.6% in March) saw maximum increases. These mixed trends were consistent even in daily O_3 concentrations after lockdown; decrease in five cities (Bangalore, Bangkok, Madrid, New York, Toronto) while increase in four (Chengdu, London, Warsaw, Wuhan) cities (Figure 9). As O_3 is a secondary air pollutant, the local availability of precursor gases (nitrogen oxides and volatile organic compounds) and local meteorology (sunlight and temperature) influences the production and destruction of O_3 in air. These two factors, particularly more incoming sunlight in March than February, are expected to vary widely across the cities and continents, and hence the O_3 concentrations too.

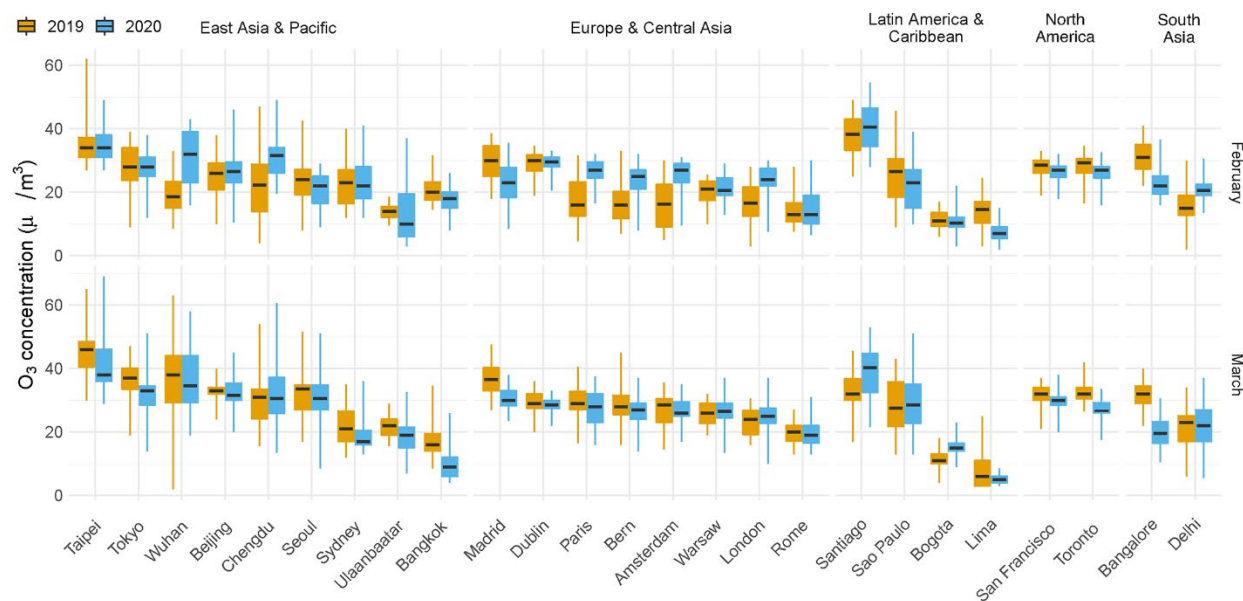


Figure 8. The box and whisker plots of daily average O₃ concentrations during February-March 2020 and 2019. The top and bottom of each box represents 75th percentile and 25th percentile respectively and the upper and lower whisker represents 90th percentile and 10th percentile respectively. The horizontal bar in each box represents the median.

There is also a mixed trend in SO₂ concentrations across cities; seven cities saw lower monthly mean SO₂ concentration in February 2020 compared to February 2019 and 12 cities have declining SO₂ in March 2020 compared to March 2019 (Figure 10). In contrast, four cities (Bangalore, Istanbul, Santiago, Ulaanbaatar) in February 2020 and five cities (Istanbul, London, Santiago, Ulaanbaatar, Toronto) in March 2020 had higher SO₂ concentrations. Maximum decline in SO₂ was observed in Lima (-75.2%), followed by Nanjing (-38.9%) and Beijing (-36.8%) in February 2020 and Bogota (-64.8%) observed a maximum decline followed by Lima (-60.4%) and Nanjing (-53.3%) in March 2020. The daily SO₂ concentration also showed mixed trends; after the lockdown, statistically significant decrease in five cities (Beijing, Delhi, Madrid, Nanjing, Vienna) and increase in 4 cities (Buenos Aires, London, Toronto, Ulaanbaatar) (Figure 11).

Mean monthly CO concentrations of February 2020 in Amsterdam (-35.1%), London (-48.3%), Lima (-27.3%), Nanjing (23.8%), Seoul (-13.4%), Chengdu (-29.1%) were lower while higher in Bangalore (+14.8%), Bogota (+15.8%), Istanbul (+171.0%), Jerusalem (+465.2%), San Francisco (+22.6%), Taipei (24.3%), Warsaw (+28.1%) than that of February 2019 (Figure 12).

In March 2020, CO concentrations decreased in 14 cities (Bangalore, Beijing, Bogota, Chengdu, Delhi, Lima, London, Nanjing, New York, San Francisco, Sao Paulo, Seoul, Toronto, Warsaw) while consistently increased in Istanbul. Daily CO concentration has significantly declined in seven cities namely (Bangalore, Bogota, Chengdu, Delhi, Lima, Nanjing, San Francisco, Sao Paulo) after the lockdown (Figure 13).

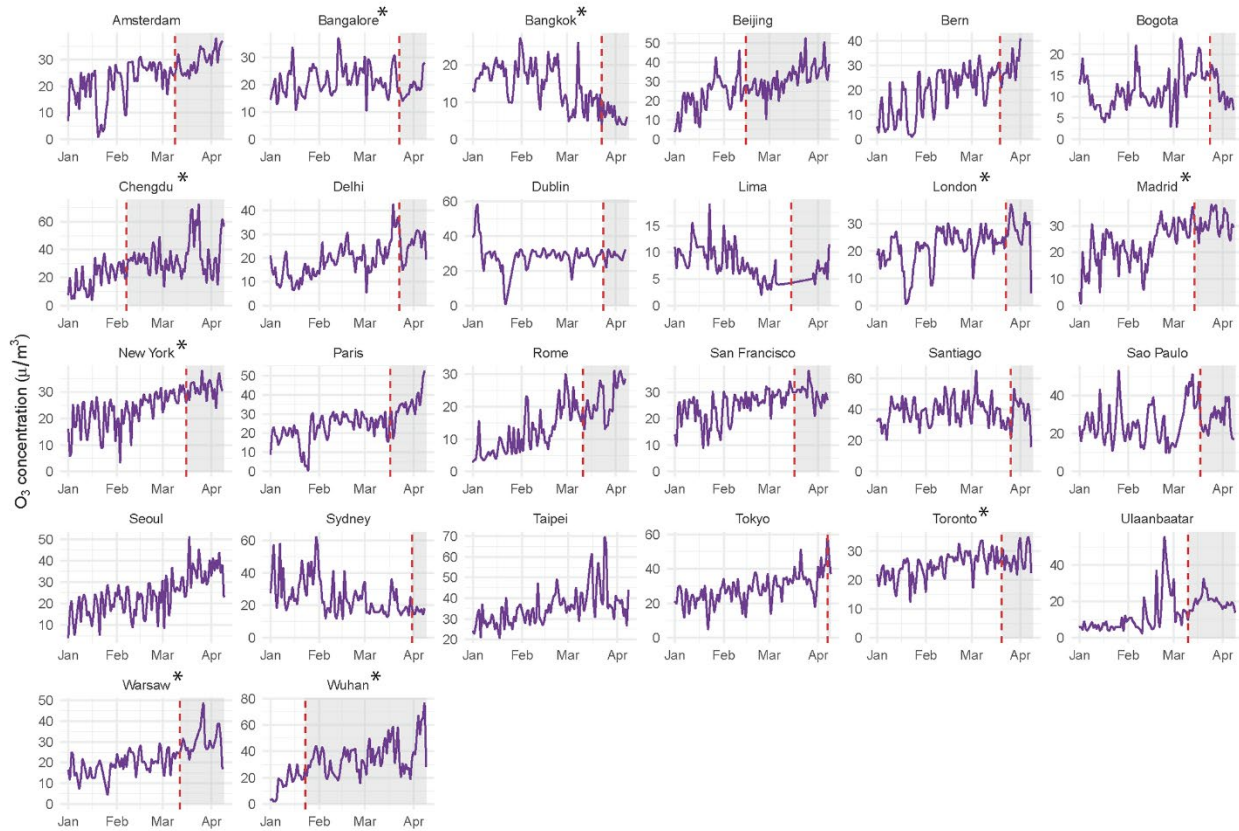


Figure 9. Time series of daily O₃ concentrations from January 1 to April 9, 2020 in 26 cities across the globe. Vertical red dashed line shows the lockdown date and * denotes significant linear trends.

Several other factors could have caused the mixed trends of pollutants such as O₃ and SO₂. First, the variation in the source of pollutants; some of the major sources of SO₂ are big coal-fired power plants that might not have been shut down during lockdown and hence SO₂ concentrations in the city would not have gone down. The seasonal forest fires in South Asia and Southeast

Asia may have also influenced the air pollutant concentrations in those cities. Second, local weather can significantly affect pollution levels—higher wind speeds tend to dilute air pollutants and rainfall can wash out pollutants from air. In some cities like Delhi and Bangkok, road vehicles are a major source of PM_{2.5}. Additionally, the high pollution level in early March prior to lockdown could have kept the monthly mean values of March this year in some cities unchanged even after shutting down traffic, businesses, and industries. In these cities, sharp decline in levels of pollutant due to lockdown is more noticeable. However, in some European cities where the sources of pollution are more varied such as from residential and commercial buildings, industrial activities, and road traffic, the changes are less apparent as such emissions could have continued despite lockdown. Furthermore, even if a city has a high contribution of the pollutant from the transport, but if the selected stations are away from roads—the impact of lockdown would be less noticeable. Emissions from natural sources such as forest fires and dust are expected to make a small change in concentrations observed in the cities unless the cities are located in or near to dominant natural sources like deserts and coasts.

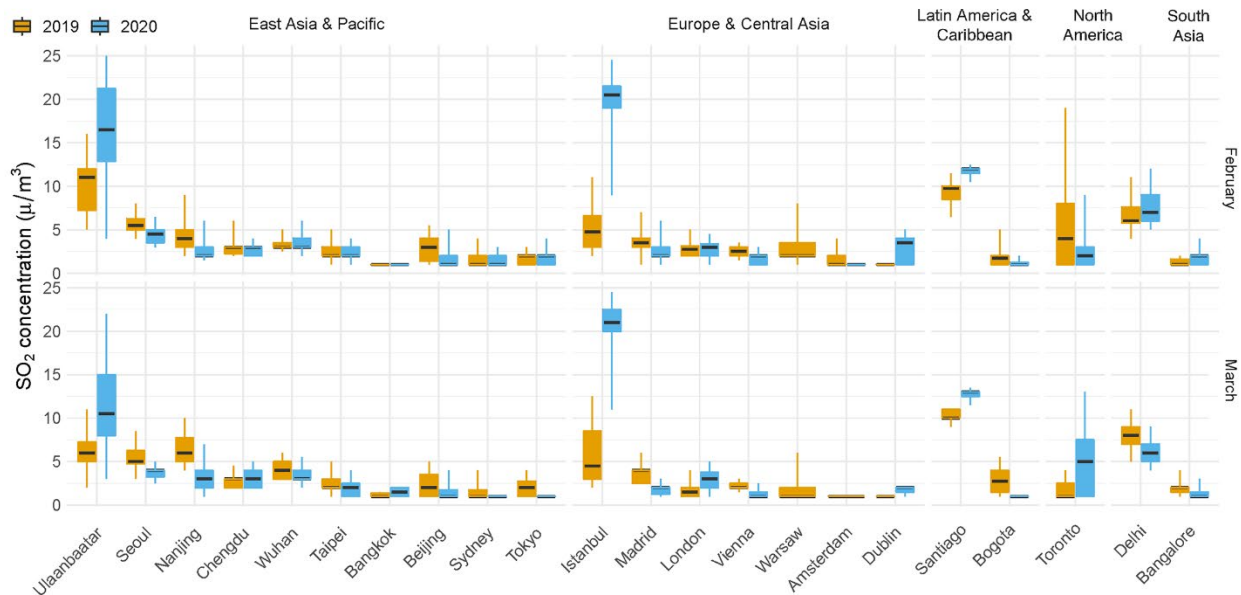


Figure 10. The box and whisker plots of daily average SO₂ concentrations during February-March 2020 and 2019. The top and bottom of each box represents 75th percentile and 25th percentile respectively and the upper and lower whisker represents 90th percentile and 10th percentile respectively. The horizontal bar in each box represents the median.

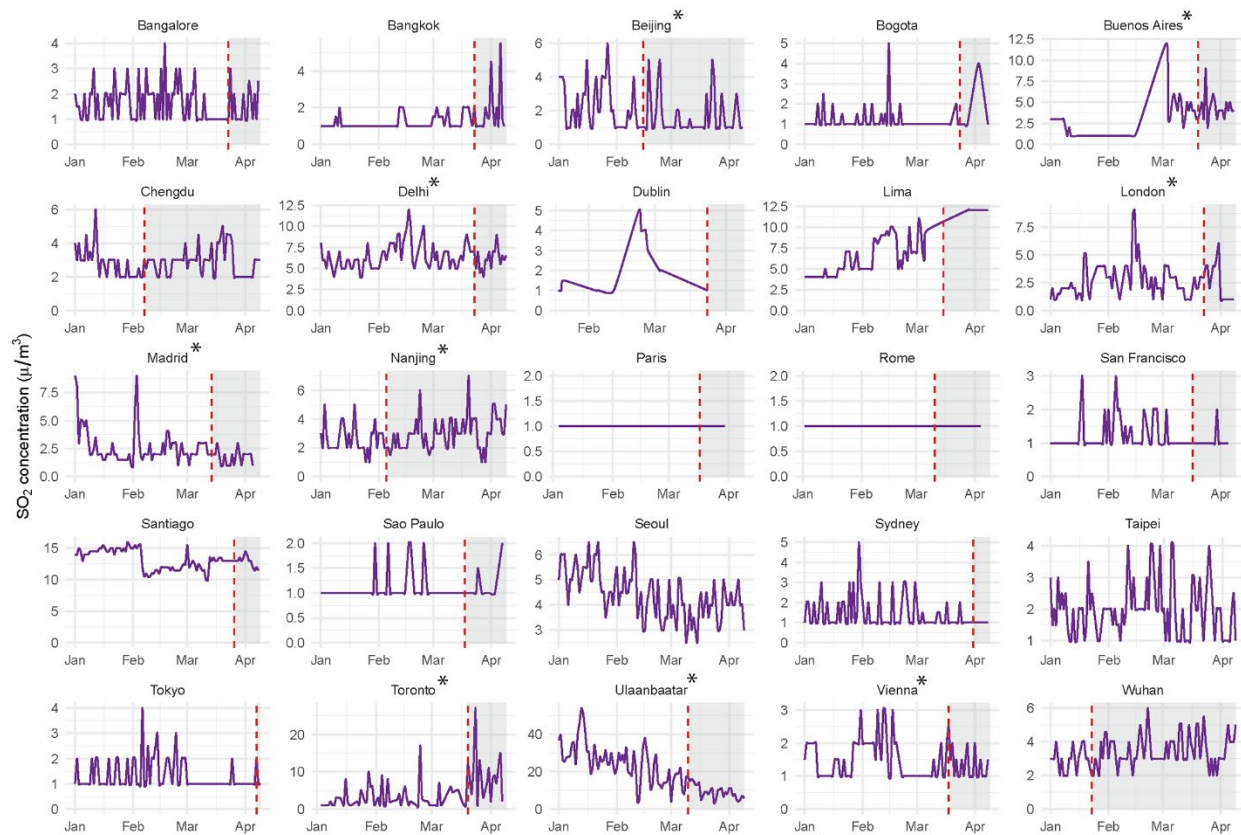


Figure 11. Time series of daily SO₂ concentrations from January 1 to April 9, 2020 in 25 cities across the globe. Vertical red dashed line shows the lockdown date and * denotes significant linear trends.

We observed consistent decrease in particulate matters (PM_{2.5}, PM₁₀) and NO₂ in cities like Bangalore, Beijing, Bern, Delhi, Lima, London, Madrid, New York, Paris, Seoul, Sydney, Tokyo, Ulaanbaatar, Vienna. Ozone (O₃) increased primarily in developed cities like Amsterdam, Bern, London, Madrid, New York, and Paris with few exceptions such as in Delhi, Wuhan, and Ulaanbaatar. A visible reduction of air pollution remains to be seen in cities like Addis Ababa and Ho Chi Minh City where the lockdowns were imposed in late March-early April. Evidence based on ground level stations shows reductions in concentrations of four major air pollutants PM_{2.5}, PM₁₀, NO₂ and CO in majority of the world cities due to almost simultaneous lockdowns. As the higher resolution data becomes available in the future, in-depth analysis considering other factors such as influence of weather should be conducted to generate more precise estimates of this impact. Moreover, further investigations are required to quantify

proportional contributions of emission reduction versus changes in meteorological conditions, and other benefits; health benefits such as lives saved due to, benefits to crops, and climate benefits (e.g., masking of global warming due to aerosols, change in aerosol-induced atmospheric warming, surface dimming or brightening) due to changes in air quality before and after lockdown.

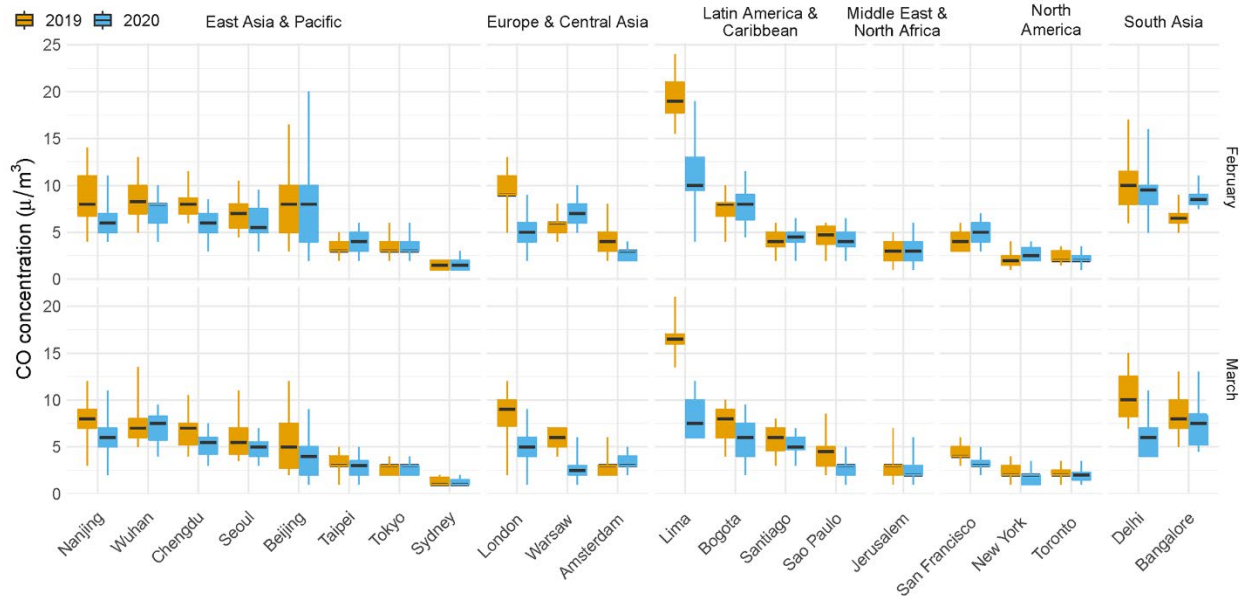


Figure 12. The box and whisker plots of daily average CO concentrations during February-March 2020 and 2019. The top and bottom of each box represents 75th percentile and 25th percentile respectively and the upper and lower whisker represents 90th percentile and 10th percentile respectively. The horizontal bar in each box represents the median.

A number of previous studies have suggested that a swift and widespread implementation of science-based clean air solutions would result in multiple benefits such as improved public health, increased crop productivity, mitigating near-term climate change, and reduced socio-economic costs due to improved air quality (e.g., UNEP, 2019). However, a simultaneous global scale, even a regional or a national scale, implementation of such air pollution control measures is almost next to impossible due to widely varying policy priorities of the countries around the world. Lockdowns introduced to contain the COVID-19 have led to a near-simultaneous shutdown of many sources of air pollutants at a global scale and has set a natural condition to examine the planetary scale reduction in air pollution albeit temporarily, and with massive suffering and socio-economic costs.

This study provides the first insight on global scale pollution reduction caused by COVID-

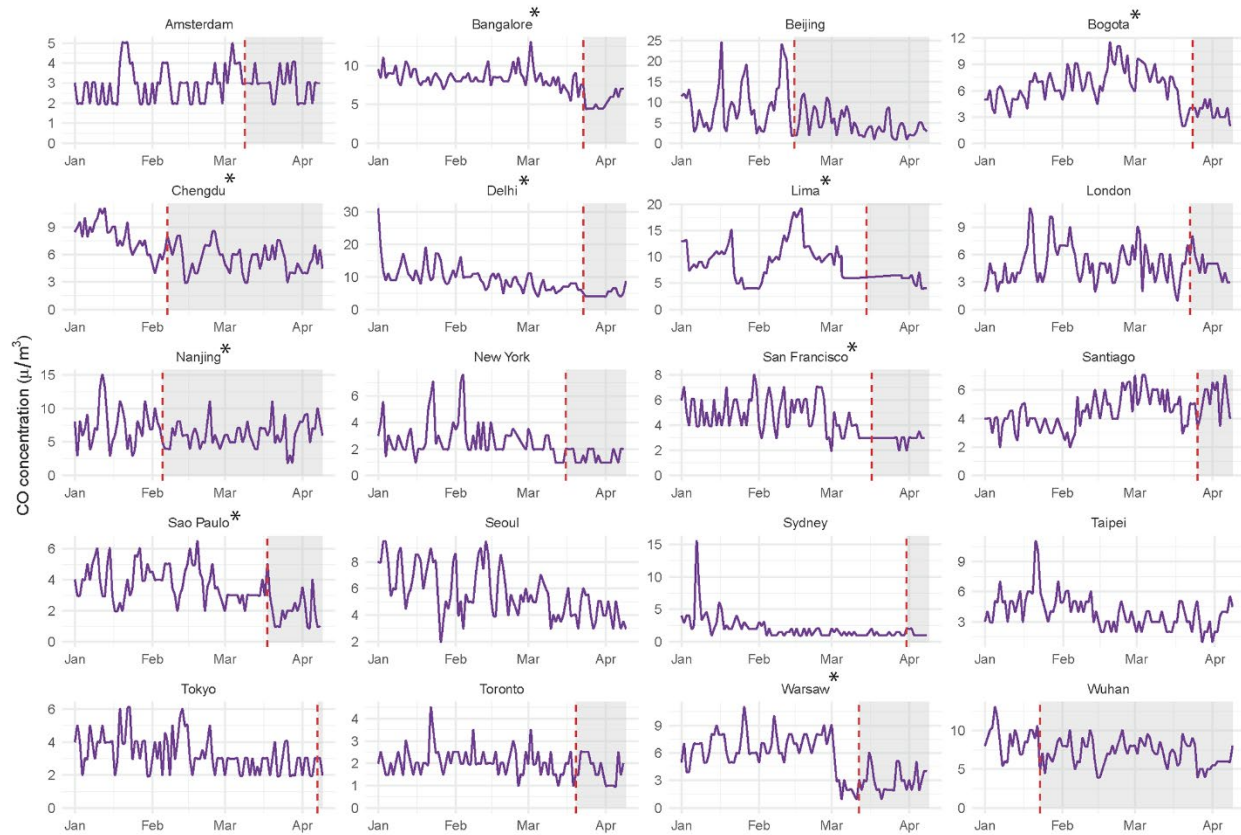


Figure 13. Time series of daily CO concentrations from January 1 to April 9, 2020 in 20 cities across the globe. Vertical red dashed line shows the lockdown date and * denotes significant linear trends.

lockdown using comprehensive air pollution data collected from 301 stations located in 40 cities in 6 continents. Though the reduction is expected to be short-lived, it provides us encouraging evidence that a swift and widespread implementation of air pollution measures, as proposed by several scientific studies, can result in immediate air quality benefits. Yet, there is nothing to celebrate as the COVID-19 pandemic has taken an enormous toll on humanity and nobody wanted air pollution reduction this way. However, it is worth to note that the similar enormous damage has been caused by air pollution globally especially in the developing countries for at least last two decades which demands actions to reduce air pollution.

The majority of the pre-existing health conditions such as respiratory and cardiovascular diseases that increase the risk of death for COVID-19 are the same diseases that are affected by long-term exposure to air pollution. In fact, air pollution aggravates the COVID-19 death rate (Wu et al. 2020). There are clear health benefits of air pollution reduction, including saving lives even during pandemic like COVID-19. The importance of continuing to enforce existing air pollution control regulations to protect human health both during and after the COVID-19 crisis cannot be and should not be ignored. While the enormous outpouring of support and global solidarity was seen that spurs numerous actions to mitigate, control, and eradicate COVID-19 pandemic, it is equally essential to show a similar level of urgency, solidarity and actions to address the great chronic environmental disasters of air pollution, a silent killer of approximately seven million people worldwide every year.

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Table 1. Description of the selected cities and news links of the lockdown information

City	Country	Region*	Income group*	Number of stations	Lockdown Start	Lockdown End	News Source
Abu Dhabi	United Arab Emirates	Middle East & North Africa	High income	1	5/04/2020	On going	https://www.outlookindia.com/outlooktraveller/travelnews/story/70181/uae-begins-24-hour-lockdown-for-two-weeks-from-april-5-to-contain-the-covid-19
Addis Ababa	Ethiopia	Sub-Saharan Africa	Low income	1	1/04/2020	On going	https://www.aljazeera.com/news/2020/04/ethiopia-declares-state-emergency-fight-covid-19-200408142519485.html
Amsterdam	Netherlands	Europe & Central Asia	High income	5	9/03/2020	On going	https://metro.co.uk/2020/03/19/trip-amsterdam-showed-government-not-enough-protect-us-coronavirus-12419537/
Bangalore	India	South Asia	Lower middle income	5	23/03/2020	On going	https://en.wikipedia.org/wiki/2020_coronavirus_pandemic_in_India
Bangkok	Thailand	East Asia & Pacific	Upper middle income	4	23/03/2020	On going	https://www.ctvnews.ca/health/coronavirus/thailand-imposes-state-of-emergency-to-control-coronavirus-1.4865483
Beijing	China	East Asia & Pacific	Upper middle income	26	15/02/2020	24/03/2020	https://www.forbes.com/sites/siminamistreanu/2020/03/27/beijing-has-defeated-coronavirus-what-it-was-like-living-in-the-locked-down-city-for-two-months/#
Bern	Switzerland	Europe & Central Asia	High income	1	19/03/2020	On going	https://www.swissinfo.ch/eng/covid-19_bern-in-the-time-of-coronavirus/45625916
Bogota	Colombia	Latin America & Caribbean	Upper middle income	14	24/03/2020	On going	https://www.aa.com.tr/en/americas/national-lockdown-starts-in-colombia-amid-protests/1779226
Buenos Aires	Argentina	Latin America & Caribbean	Upper middle income	1	20/03/2020	On going	https://www.latinfinance.com/daily-briefs/2020/3/31/argentina-braces-for-gradual-economic-recovery-after-lockdown

Chengdu	China	East Asia & Pacific	Upper middle income	10	2/07/2020	On going	http://society.people.com.cn/n1/2020/0207/c1008-31576755.html
Delhi	India	South Asia	Lower middle income	14	23/03/2020	On going	https://en.wikipedia.org/wiki/2020_coronavirus_pandemic_in_India
Dhaka	Bangladesh	South Asia	Lower middle income	1	26/03/2020	On going	https://www.aa.com.tr/en/asia-pacific/covid-19-bangladesh-extends-shutdown-until-april-11/1786610
Dublin	Ireland	Europe & Central Asia	High income	2	24/03/2020	On going	https://en.wikipedia.org/wiki/2020_coronavirus_pandemic_in_the_Republic_of_Ireland#COVID-related_offences
Ho Chi Minh City	Vietnam	East Asia & Pacific	Lower middle income	1	31/03/2020	On going	https://www.voanews.com/science-health/coronavirus-outbreak/vietnam-latest-lockdown-while-us-braces-huge-death-toll
Istanbul	Turkey	Europe & Central Asia	Upper middle income	4	2/04/2020	On going	https://en.qantara.de/content/turkey-mayor-urges-lockdown-for-istanbul
Jakarta	Indonesia	East Asia & Pacific	Lower middle income	3	10/04/2020	On going	https://www.thestar.com.my/news/regional/2020/04/08/jakarta-to-impose-partial-lockdown-on-april-10
Jerusalem	Israel	Middle East & North Africa	High income	3	7/04/2020	On going	https://www.i24news.tv/en/news/israel/1586192909-israel-s-netanyahu-announces-new-anti-coronavirus-restrictions
Kampala	Uganda	Sub-Saharan Africa	Low income	1	30/03/2020	On going	https://www.kfm.co.ug/entertainment/photos-empty-streets-of-kampala-after-lockdown.html
Kathmandu	Nepal	South Asia	Low income	4	24/03/2020	On going	https://nep123.com/covid-19-kathmandu-valley-wears-a-disserted-look-with-lockdown-enforcement/
Lima	Peru	Latin America & Caribbean	Upper middle income	11	15/03/2020	On going	https://www.miningmagazine.com/covid-19/news/1383191/covid-19-lock-down-affects-miners-in-peru
London	United Kingdom	Europe & Central Asia	High income	20	23/03/2020	On going	https://www.nytimes.com/2020/03/24/world/europe/britain-coronavirus-lockdown.html
Madrid	Spain	Europe & Central Asia	High income	4	14/03/2020	On going	https://en.as.com/en/2020/03/26/other_sports/1585233133_070352.html

Nanjing	China	East Asia & Pacific	Upper middle income	3	5/02/2020	On going	https://en.wikipedia.org/wiki/2020_Hubei_lockdowns
New York	United States	North America	High income	12	16/03/2020	On going	https://news.sky.com/story/coronavirus-new-york-and-la-on-lockdown-as-world-reacts-to-spread-of-covid-19-11958145
Oslo	Norway	Europe & Central Asia	High income	3	12/03/2020	On going	https://en.wikipedia.org/wiki/2020_coronavirus_pandemic_in_Norway
Paris	France	Europe & Central Asia	High income	11	17/03/2020	On going	https://www.aa.com.tr/en/europe/parisians-vacate-city-ahead-of-covid-19-lockdown/1769664
Rome	Italy	Europe & Central Asia	High income	10	11/03/2020	On going	https://www.repubblica.it/politica/2020/03/11/news/coronavirus_conte_italia_governo_misure-250988471/?ref=RHPPTP-BL-I250988111-C12-P1-S1.12-T1
San Francisco	United States	North America	High income	7	17/03/2020	On going	https://www.aa.com.tr/en/americas/us-san-francisco-under-3-week-coronavirus-lockdown/1768830
Santiago	Chile	Latin America & Caribbean	High income	6	26/03/2020	On going	https://en.wikipedia.org/wiki/2020_coronavirus_pandemic_in_Chile#Government_response
Sao Paulo	Brazil	Latin America & Caribbean	Upper middle income	14	18/03/2020	On going	https://www.aa.com.tr/en/americas/death-toll-in-brazil-from-covid-19-rises-to-159/1785792
Seoul	South Korea	East Asia & Pacific	High income	6	No lockdown	On going	
Sydney	Australia	East Asia & Pacific	High income	17	31/03/2020	On going	https://www.theguardian.com/australia-news/2020/apr/01/social-distancing-rules-australia-lockdown-coronavirus-strict-new-laws-legal-illegal-state-physical-restrictions-guidelines-explained-nsw-victoria-qld-queensland-act-sa-wa-two-person
Taipei	Taiwan	East Asia & Pacific	High income	1	No lockdown	On going	
Tashkent	Uzbekistan	Europe & Central Asia	Lower middle income	1	24/03/2020	On going	http://www.china.org.cn/world/Off_the_Wire/2020-03/23/content_75850320.htm

Tokyo	Japan	East Asia & Pacific	High income	20	7/04/2020	On going	https://edition.cnn.com/2020/04/07/asia/japan-coronavirus-state-of-emergency-intl-hnk/index.html
Toronto	Canada	North America	High income	12	20/03/2020	On going	https://www.cbc.ca/news/canada/toronto/coronavirus-covid-19-ontario-friday-online-learning-1.5504109
Ulaanbaatar	Mongolia	East Asia & Pacific	Lower middle income	6	10/03/2020	On going	https://www.garda.com/crisis24/news-alerts/321181/mongolia-government-places-ulaanbaatar-and-other-cities-on-lockdown-due-to-covid-19-march-10-update-7
Vienna	Austria	Europe & Central Asia	High income	12	18/03/2020	On going	https://metropole.at/coronavirus-in-austria/
Warsaw	Poland	Europe & Central Asia	High income	3	12/03/2020	On going	https://en.wikipedia.org/wiki/2020_coronavirus_pandemic_in_Poland
Wuhan	China	East Asia & Pacific	Upper middle income	21	23/01/2020	24/03/2020	https://news.cgtn.com/news/2020-04-01/Study-China-lockdown-may-have-blocked-700-000-COVID-19-cases-PkzcYgVj32/index.html

Country specific classification of income group and region were adopted from the World Bank (<https://blogs.worldbank.org/opendata/new-country-classifications-income-level-2019-2020>, accessed April 7, 2020).
