

A Study of Impact of Air Pollutants on Human Health during COVID-19 in Delhi

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Abstract: *The aim of the study is to envisage the dose of deposition of particulate matter mainly PM_{2.5} and PM₁₀. 35 monitoring stations as installed by Central Pollution Control Board have been considered under the study. International Commission on Radiological Protection method also known as RDD estimation was used in investigating the Station wise value of RDD_{2.5} and RDD₁₀ during standing and walking during. Apart from the empirical study, GIS and remote sensing techniques also used to see the effect of population density. The results show that the values of RDD₁₀ during standing and walking were found to be relatively more in dense populated areas, however the values of RDD_{2.5} found to be more in relatively less dense populated areas. Therefore, the deposition of particulate matter which played an important role in the enhancement of mortalities during Novel Corona Virus occurred in dense populated areas. The GIS study also shows that on the basis of pollution data available from CPCB, 250 no. of people per kilo-meter square are under exposure of PM_{2.5} and PM₁₀.*

Keywords: Particulate Matter, RDD, DF, Precipitation, Buffer, Zonal statistics, Arc GIS etc.

1. Introduction

Air pollution refers to the release of various gases and finely divided solids or dispersed aerosols that disturb the natural state of the air. Air pollution now become a major issue to human and environmental health worldwide [1]. Worldwide every year, 7 million premature deaths are due to particulate matter (PM) [2]. In India 1.67 million deaths were reported to air pollution in 2019, resulting in 17.8% of the total deaths in the country. The main causes of these deaths were outdoor particulate matter and indoor air pollution [3]. PM is a suspended particle composed of liquid and solids in air derived from different sources and is present in different sizes [7-8]. PM can be classified into two categories, pollutants which are directly emitted from the source known as primary pollutants and secondary pollutants which are formed in the chemical reaction of sulfur dioxide and nitrogen oxide which are emitted from industries, power plants, and automobiles [6-7]. Concentration of PM depend on population density [6]

Particulate matter is of three types coarse particle PM₁₀ (particle having a diameter less than 10 μ m), fine particle PM_{2.5} (particle having a diameter less than 2.5 μ m) and ultrafine particle PM_{0.1} (particle having a diameter less than or 0.1 μ m).

According to past experimental and empirical studies, PM has shown negative impacts on human health [9-10][9] and falls under the category of group 1 carcinogen to humans. According to several studies, it has been confirmed that PM₁₀ harm human health and can result in medical conditions such as congenital heart defects [10], ischemic heart disease [11], respiratory and circulatory mortality [12], preterm-birth risk [13], mutagenicity and DNA damage [14], fetal growth characteristics and adverse birth outcomes [15], inflammatory

responses [16] and increased cancer risk [17]. Apart from this, PM also reduces visibility and harms air quality [19-20].

In Wuhan, China on December 2019 the first novel coronavirus was reported [18]. World Health Organization declared COVID-19 as a pandemic on 11 March 2020 as it was spreading worldwide. COVID-19 virus belongs to the *coronaviridae* family which causes a variety of respiratory diseases in human beings [18]. It was found that until 5 December 2022, 4.6 million cases and 17 thousand deaths were reported in china. Till December 2022, 4.4 million cases and 5 lakh deaths were reported in India [19]. Various measures were taken by the Indian government to control the situation by imposing a series of lockdowns with concomitant restrictions on commercial, industrial and social activities. The first complete nationwide lockdown was introduced on 24th March 2020 to 14th April 2020 when all the services like institutional, commercial, and industrial were suspended including ban on the movement of citizen except essential services such as medical service police service. The nationwide lockdown was extended from 15 April 2020 – 3 May 2020. The third lockdown was imposed from 4 May 2020 – 17 May 2020 dividing the country into three zones red (area which have high number coronavirus cases and it was remain under lockdown), orange (area which have comparatively less cases than red zone and only private vehicle were allowed with no public transportation), and green zones (area which do not have any corona cases in the past 21 days and normal movement was allowed with 50 percent bus capacity). Forth lockdown was imposed from 18 May 2020 -31 May 2020 while further dividing the red zone into contaminated and buffer zone. From 1 June 2020 – 31 June 2020 unlock phase 1 was imposed with the opening up of the shopping malls, religious places, hotels, and restaurants with gathering of up to

50 people were allowed while the night curfew (9:00 PM – 5:00 AM) was still there.

According to a study, there is a strong correlation between the oxidative potential of PM, genotoxicity, ecotoxicity and PM's exposure [20]. Exposure to PM for a longer period has an impact on aggravating emerging diseases because of the presence of Severe Acute Respiratory Syndrome (SARS) and COVID-19 variants in the atmosphere [21]. According to research carried out in Italy, it was found that the major cause of COVID-19 was air pollution to human transmission rather than human-to-human transmission [23-24].

International commission on radiological protection (ICRP) and the National Council on Radiation Protection and Measurement (NCRP) are the two Mathematical models were used for the prediction of the total and regional deposition of particles. For assessment of health risks due to airborne particulate matter, Respiratory lung deposition dose is determined by calculating the amount of particles that are deposited in lungs per unit of time. These values are calculated based on the ICRP model used in different physical activities [24]. According to past studies walking/exercising near busy road lead to more PM deposition than working out in parks [25]. Previous studies have shown that particulate matter is deposited more during walking/exercise than rest [26] as breathing rate is found to increase during physical activity which leads to an increase in the concentration of PM as it is directly related to breathing rates. [28-29].

Delhi is one of the most polluted cities in the world [28] It has a higher level of PM in the ambient atmosphere including fine ($PM_{2.5}$) and coarse (PM_{10}). During COVID 19 India went through different lockdown phases which result in variations of pollutants due to a decrease in industrial activity, and lesser/no transportation. Delhi comes in the list of most polluted cities in the world, and the sudden change of PM concentration and its deposition was studied. The present study uses the RDD concentration of both fine and coarse particles at the different sites during, before, and after lockdown phases in Delhi, India. As RDD is directly connected to PM concentration

2. Materials & Methods/ Methodology

Site Description

In the present study, Delhi has been chosen as a study area where the Central Pollution Control Board, India (CPCB) Delhi has installed 40 stations, out of which 35 locations are considered as 5 station's data was unavailable. The whole city is divided into 11 areas as administrative area, institutional area, residential area, institutional-residential area, commercial-administrative area, village area, residential-industrial area, industrial area, commercial area, institutional-commercial area and residential-commercial area describes in Figure 1.

Data Collection

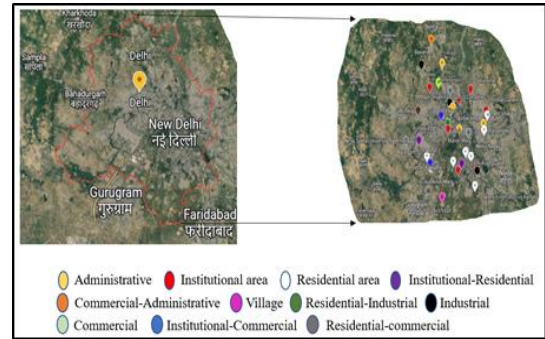


Figure 1: Location of Air Quality Monitoring station and their locations

In the present study real-time data have been collected from the monitoring stations of the central pollution control board (CPCB), and Delhi pollution control board (DPCC) of $PM_{2.5}$ and PM_{10} concentration ($\mu g/m^3$). All the raw data used in this study is available at the CPCB portal (<https://cpcb.nic.in/automatic-monitoring-data/>). Data from 24 January 2020 to 30 June 2020 have been studied to analyses air quality before, during and after the lockdown COVID-19 phase.

The air quality data has been divided into 6 parts, corresponding to the lockdown and unlock phases in the city of Delhi. More details on the differences between these phases and timelines are detailed in Table 1

Data Analysis

In the present study, International Commission on Radiological Protection method also known as RDD estimation is being used [29]. In this study for RDD calculation, three equations are used [30]–[32].

$$RDD \text{ of } PM = (V_T * f) * DF * PM \quad (1)$$

Where V_T stands for tidal volume (m^3 per breath), f represents typical breathing frequency (breath per minute), and DF shows deposition fraction of a size fraction i .

PM_i = Mass concentration in different size ranges (in $\mu g/m^3$)

$$DF = IF \left(0.058 + \frac{0.911}{1 + \exp(4.77 + 1.485 \ln dp)} + \frac{0.943}{1 + \exp(0.508 - 2.58 \ln dp)} \right) \quad (2)$$

$$IF = 1 - 0.5 \left(1 - \frac{1}{1 + 0.00076 d_p^{2.8}} \right) \quad (3)$$

The V_T and f values mainly depend on the physical activity and gender of a person [30], [32]. In the present study, to evaluate RDD, all the calculations were done using values available for males to know the general trend of RDD. For walking mode V_T and f values used here are $12.5 \times 10^{-4} m^3$ per breath and 20 breaths per minute, respectively and for sitting mode $7.5 \times 10^{-4} m^3$ per breath and 12 breaths per minute, respectively for males. For DF_i calculation, particle diameter (represented as d_p) of $PM_{2.5}$ and PM_{10} is used here [33]. RDD_{2.5} walking and sitting represent the Respiratory lung

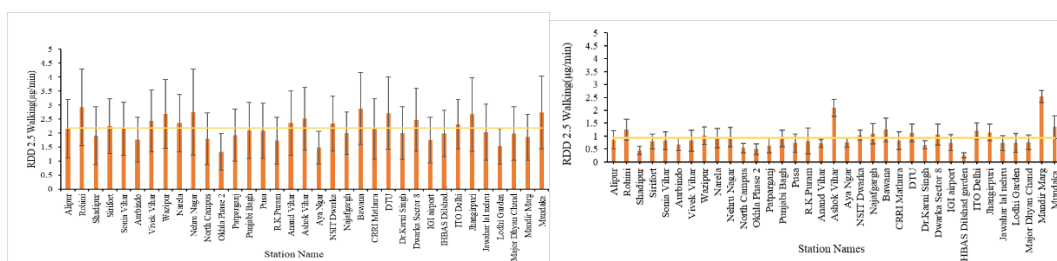
deposition dose of PM_{2.5}. RDD10 walking and sitting represent the Respiratory lung deposition dose of PM₁₀.

Table 1: Major relaxation and restriction during different lockdown phases[35]

Phase Name	Dates	Major Relaxation And Restriction
Before lockdown	24 January-24 March 2020	All activities in the usual mode, with no restrictions at all
Lockdown Phase -1	25 March-14 April 2020	All private and commercial, hospitality services, transport, and industries establishment closed, and schools, colleges, and shopping malls were closed including the ban on the movement of a citizen except for essential services such as medical and police services.
Lockdown Phase -2	15 April-3 May 2020	Farming and some industries were allowed. All restrictions were the same as in lockdown phase 1
Lockdown Phase -3	4 May-17 May 2020	Classification of cities into three zones (Orange, red and green) green and the orange zone was relaxed. Activities during phase 2 and construction, vehicle movement for some activities were permitted.
Lockdown Phase -4	18 May-31 May 2020	Permission is granted for the opening of industries and the movement of vehicles without restriction. The red zone was further divided into contaminated and buffer zone.
Unlock Phase -1	1 June-30 June 2020	All religious places, hotels, shopping malls, and hospitality services were opened. Night curfew was still there and large gatherings were still not allowed.

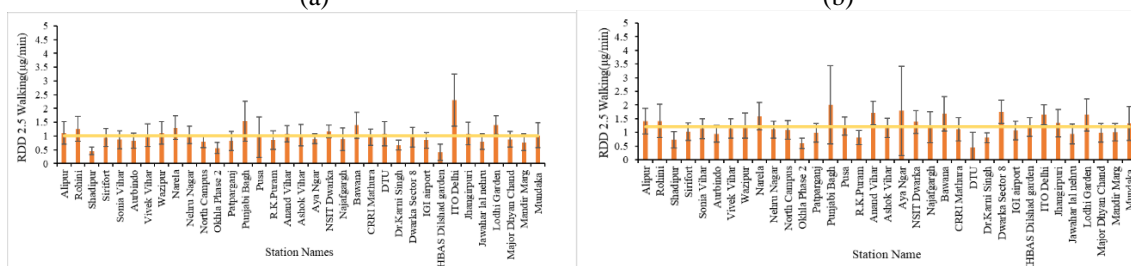
Table 2: Top three RDD values during different lockdown Phase

	RDD WALKING		RDD SITTING	
	PM _{2.5}	PM ₁₀	PM _{2.5}	PM ₁₀
Before lockdown	Rohini(2.914±1.374)	Dwarka Sector 8 (5.842±2.125)	Bawana(1.033±0.459)	Dwarka Sector 8(2.103±0.765)
	Bawana(2.87±1.578)	Wazipur (5.041±1.577)	NehruNagar(0.988±0.553)	Wazipur(1.815±0.568)
	Nehru Nagar(2.744±1.578)	DTU(5.024±1.290)	Mundaka(0.982±0.466)	DTU(1.809±0.729)
lockdown phase 1	Mandir Marg(2.535±0.238)	Najafgarh(3.04±0.917)	Ashok Vihar(0.755±0.119)	Ashok Vihar(1.236±0.198)
	Ashok Vihar(2.035±0.331)	Bawana(2.594±0.884)	Mundaka(0.487±0.157)	Najafgarh(1.094±0.33)
	Mundaka(1.353±0.437)	Rohini(2.363±0.862)	Bawana(0.447±0.164)	Rohini(0.851±0.310)
lockdown phase 2	ITO(2.305±0.953)	North campus(4.281±1.876)	Punjabi Bagh(0.523±0.306)	North Campus(1.541±0.411)
	Punjabi Bagh(1.538±0.729)	Alipur(3.673±1.2230)	Bawana(0.5±0.173)	Bawana(1.269±0.659)
	Lodhi Garden(1.405±0.334)	Rohini(2.9520.983)	Rohini(0.453±0.162)	Ashok Vihar(1.134±0.285)
lockdown phase 3	Punjabi Bagh(2.003±1.436)	Narela(3.683±1.096)	Mandir Marg(0.999±0.116)	Rohini(0.509±0.220)
	Aya Nagar(1.796±0.177)	Alipur(3.364±1.110)	Punjabi Bagh(0.721±0.517)	Mandir Marg(2.281±0.252)
	Bawana(1.389±0.481)	Punjabi bagh(3.421±1.909)	Rohini(0.509±0.220)	Narela(1.326±0.394)
lockdown phase 4	Punjabi Bagh(2.137±1.794)	Narela(4.576±1.873)	Punjabi Bagh(0.769±0.646)	Narela(1.647±0.674)
	Bawana(1.753±0.83)	DTU(4.483±1.876)	Bawana(0.631±0.299)	Alipur(1.627±0.694)
	NSIT(1.726±0.549)	Jahangirpuri(3.118±0.434)	NSIT(0.622±0.198)	Bawana(1.6±0.611)
Unlock phase 1	ITO(1.576±0.522)	Bawana(3.522±1.406)	ITO(0.567±0.188)	Bawana(1.268±0.506)
	NSIT(1.421±0.483)	Narela(3.364±1.267)	NSIT(0.511±0.174)	Dwarka Sector 8(1.161±0.476)
	Anand Vihar(1.414±0.411)	Dwarka Sector 8(3.226±1.322)	Anand Vihar(0.505±0.148)	Narela(1.211±0.459)



(a)

(b)



(c)

(d)

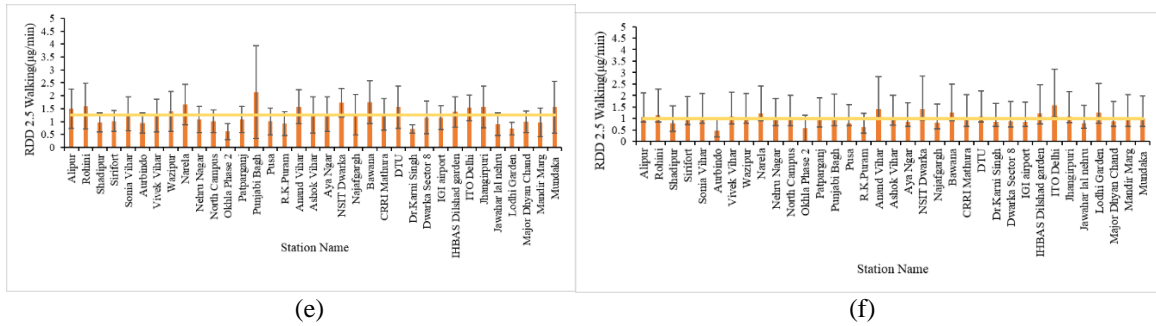


Figure 2: RDD2.5 Walking values during (a) before lockdown (b) lockdown phase 1 (c) lockdown phase 2 (d) lockdown phase 3 (e) lockdown phase 4 (f) unlock phase 1. The vertical error bars represent the standard deviation

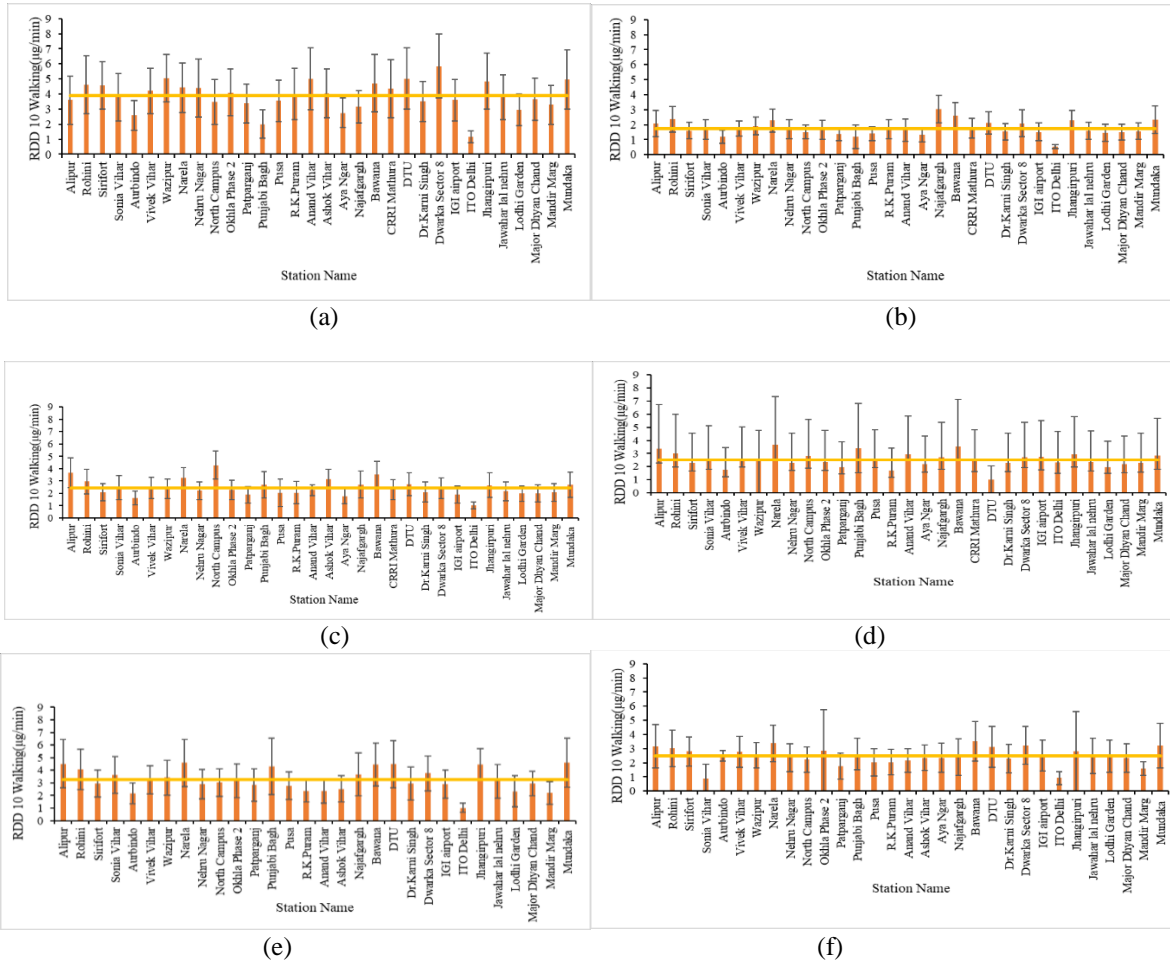
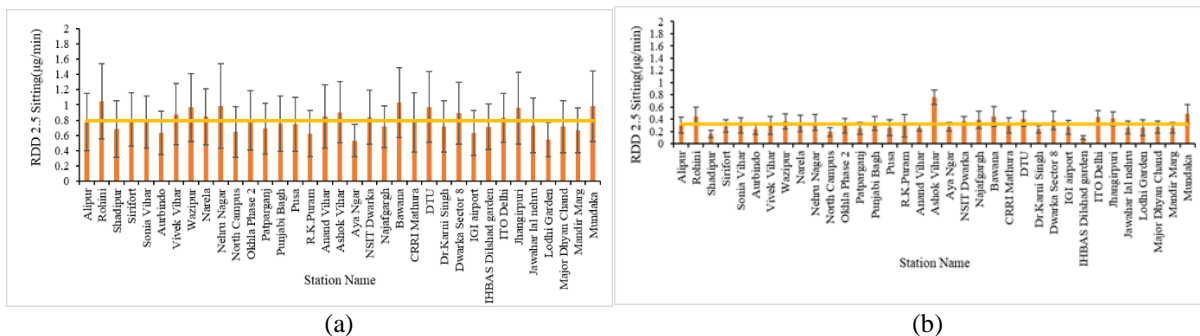


Figure 3: RDD 10 Walking values during (a) before lockdown (b) lockdown phase 1 (c) lockdown phase 2 (d) lockdown phase 3 (e) lockdown phase 4 (f) unlock phase 1. The vertical error bar represents the standard deviation.



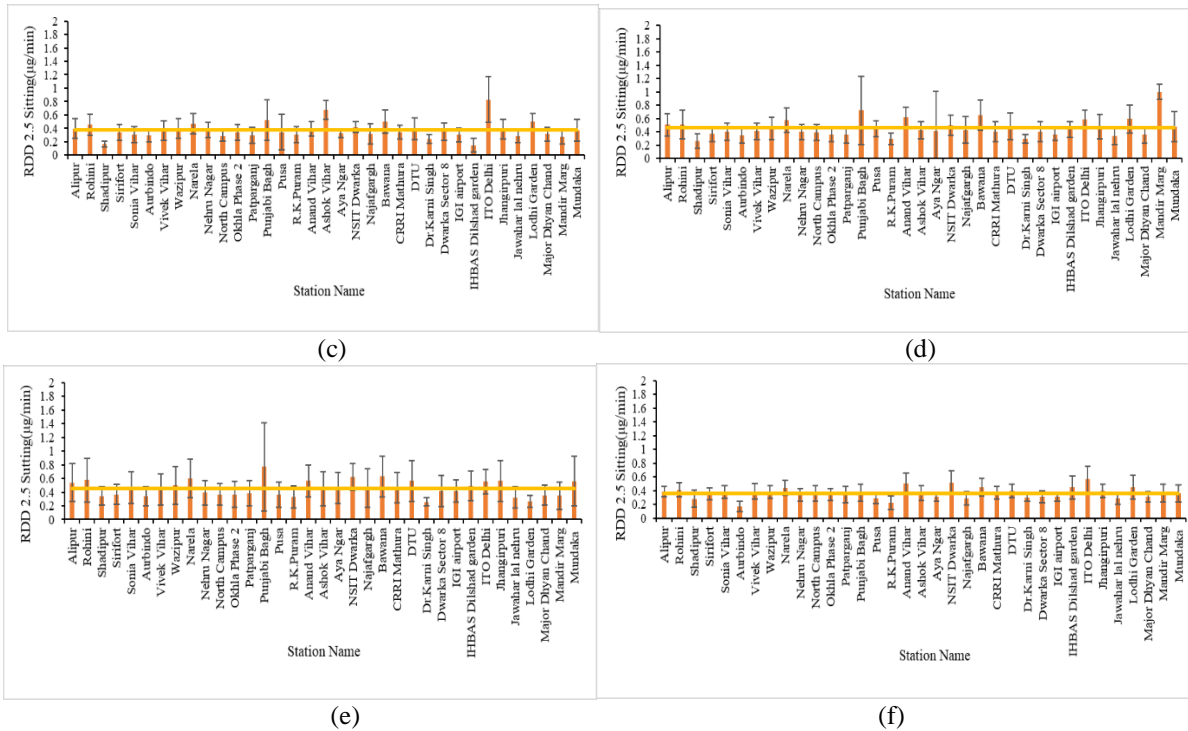
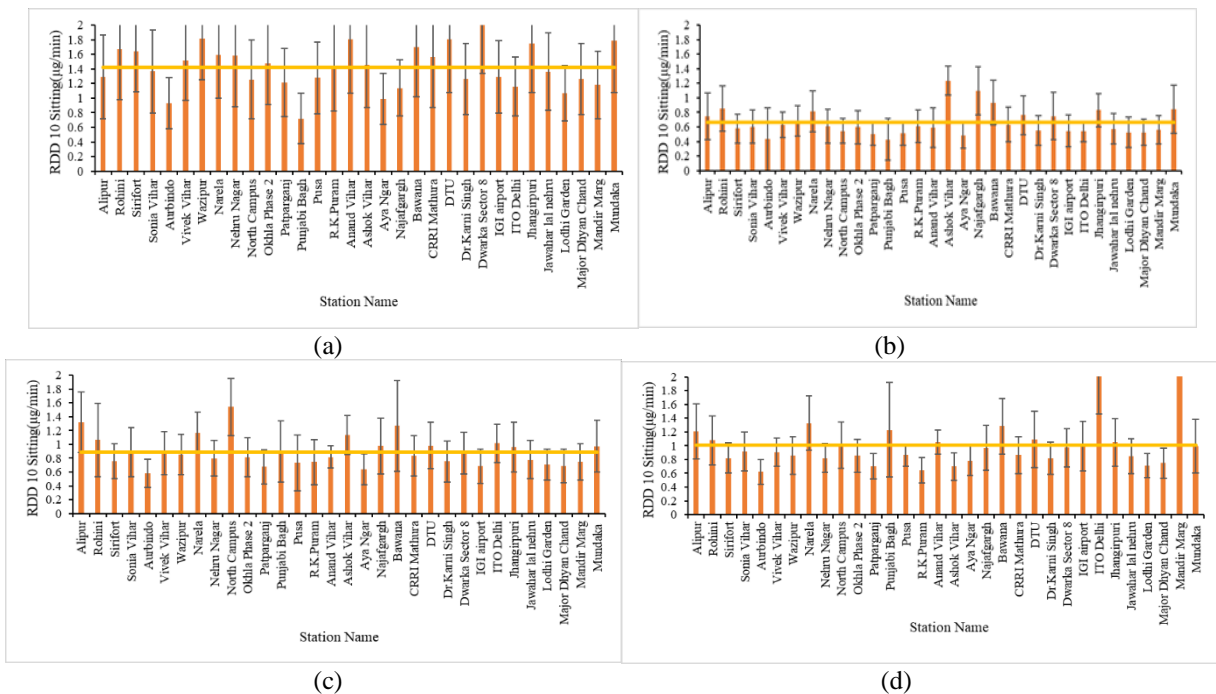
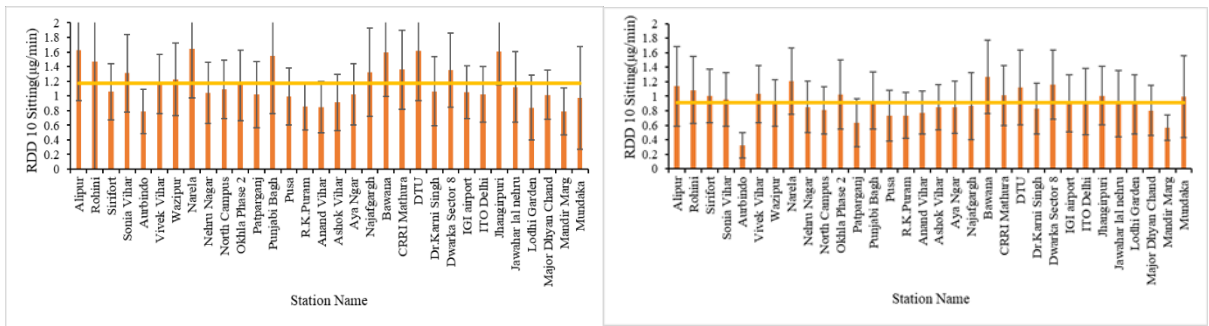


Figure 4: RDD 2.5 Sitting values during (a) before lockdown (b) lockdown phase 1 (c) lockdown phase 2 (d) lockdown phase 3 (e) lockdown phase 4 (f) unlock phase 1. The vertical error bar represents the standard deviation.





(e)

(f)

Figure 5: RDD 10 Sitting values during (a) before lockdown (b) lockdown phase 1 (c) lockdown phase 2 (d) lockdown phase 3 (e) lockdown phase 4 (f) unlock phase 1. The vertical error bar represents the standard deviation.

While conducting the GIS study, the shape file of Delhi ward boundary, 2022 has been uploaded from Environmental Systems Research Institution (ESRI). The pollution data of 35 stations have been mapped in GIS environment under Geographic co-ordinate system, GCS_WGS_1984.

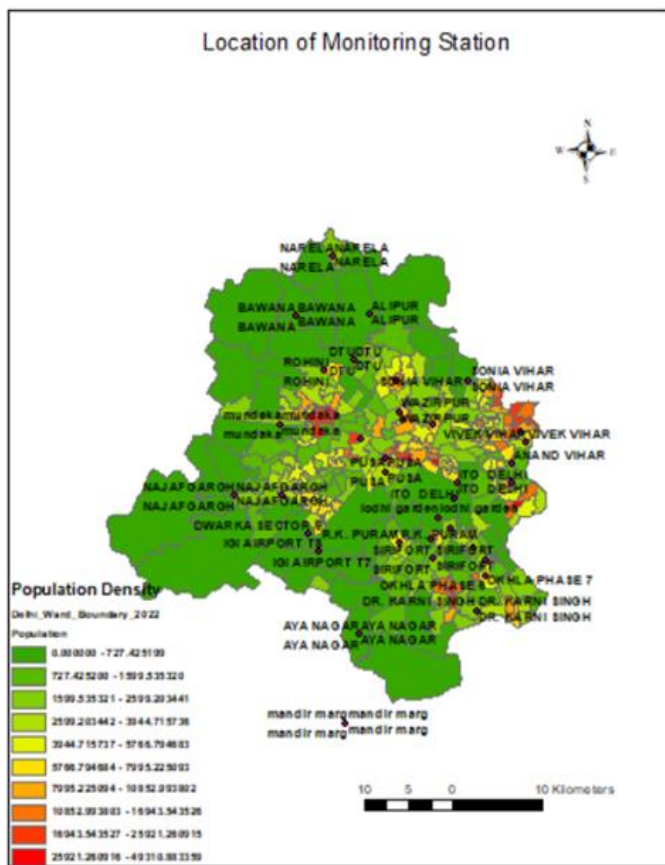


Figure 6: Location of Pollution Monitoring Station

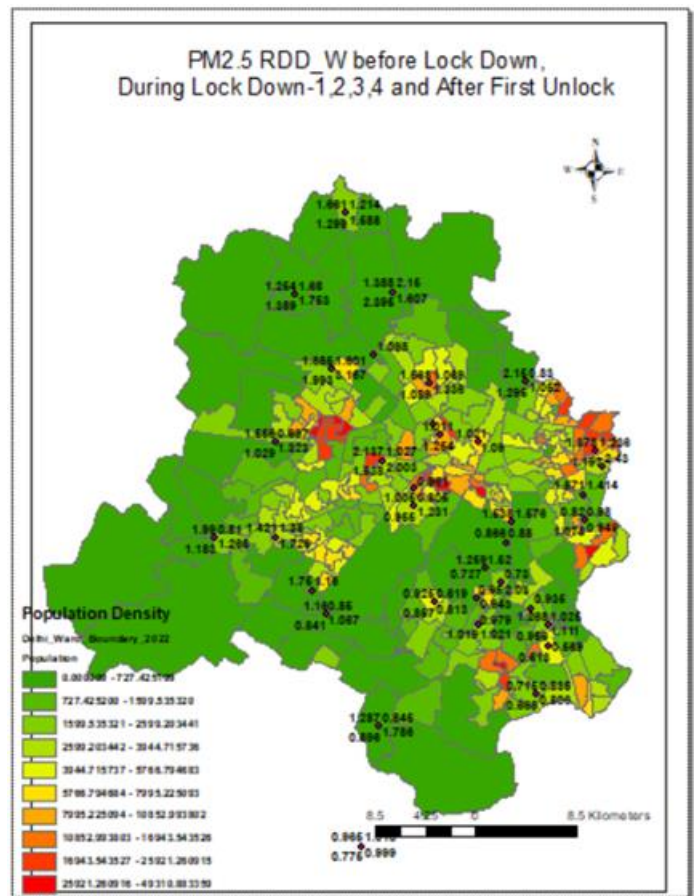


Figure 7: PM2.5 RDD_W before lock down and during lock down ph-1,2,3,4 and after first unlock

The above map shows that RDD load was higher in dense populated area except Bawana as it is industrial area, hence RDD load is more despite being non-dense populated area.

In the above map, legends show the population density, which has been reclassified in 10 sections.

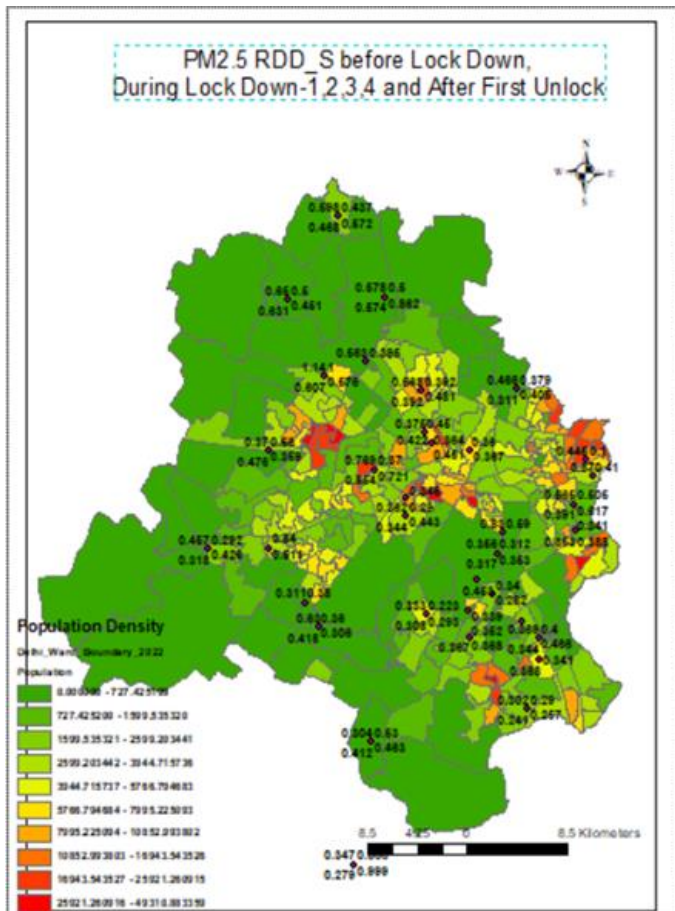


Figure 8: PM2.5 RDD_S before lock down and during lock down ph-1,2,3,4 and after first unlock

The above map shows that RDD load was again higher in dense populated area even in sitting condition except Bawana as it is industrial area, hence RDD load is more despite being non-dense populated area.

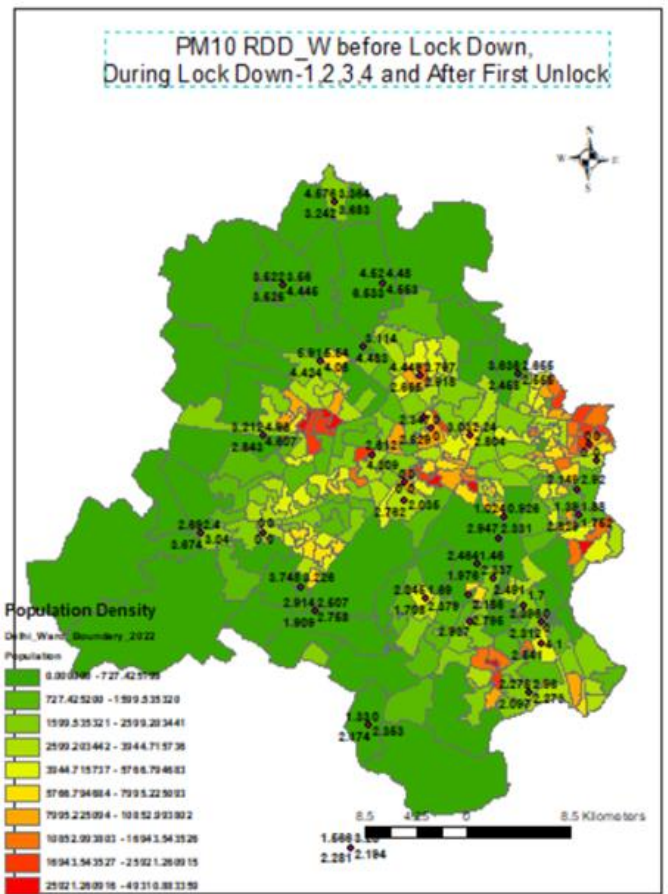


Figure 9: PM10 RDD_W before lock down and during lock down ph-1,2,3,4 and after first unlock

The above map shows that RDD load is higher in dense populated area except Bawana as it is industrial area, hence RDD load is more despite being non-dense populated area.

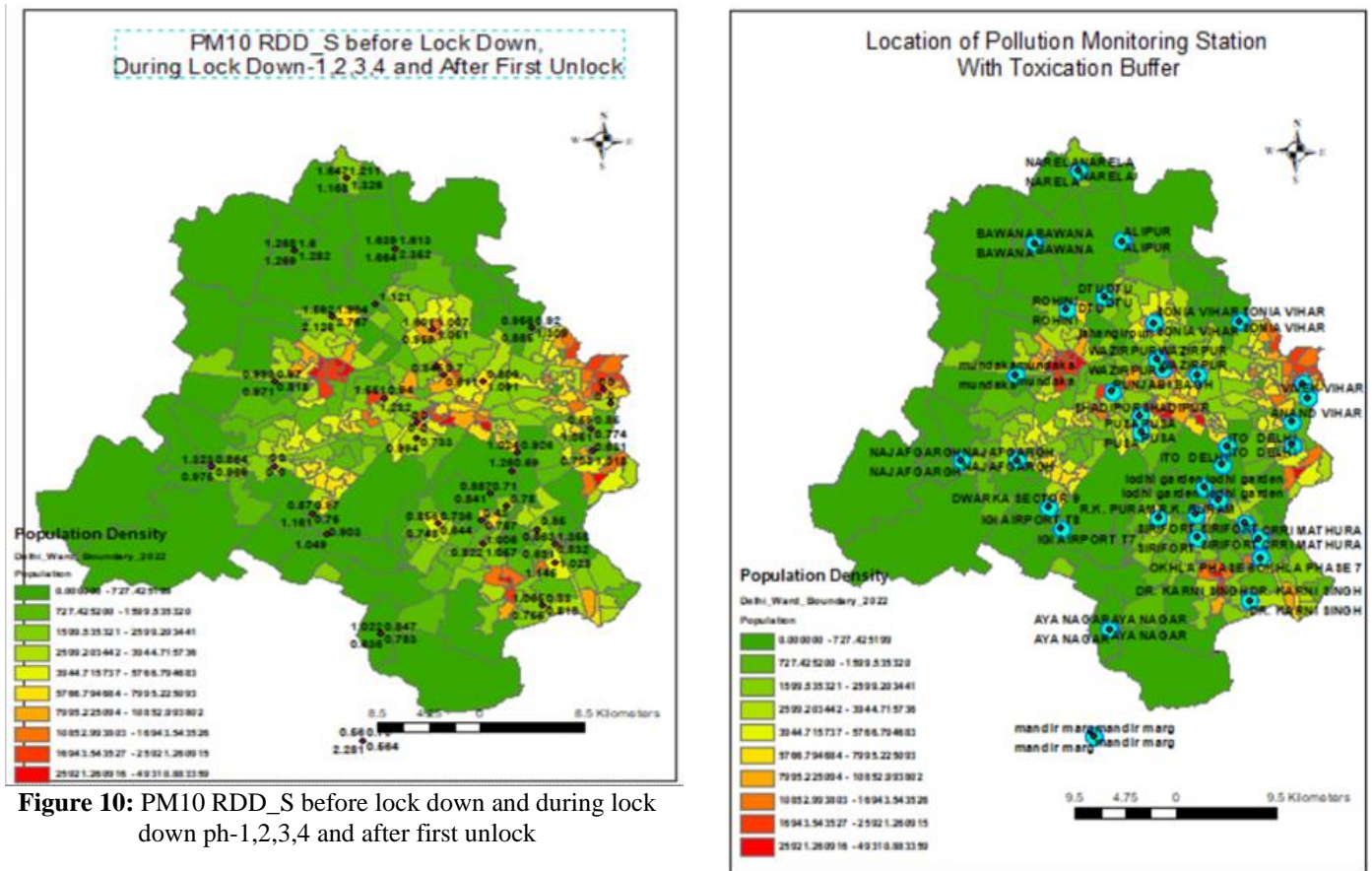
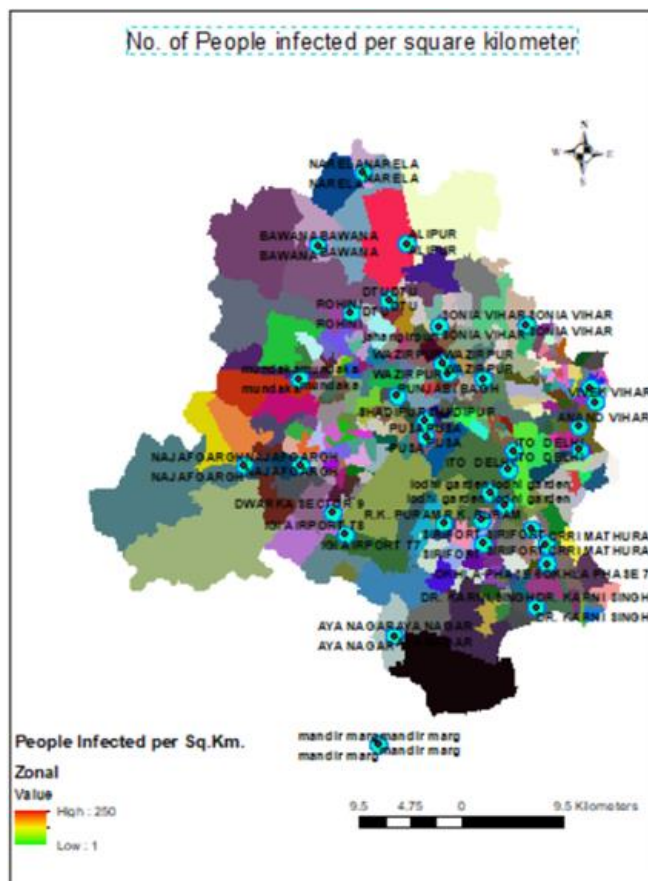


Figure 10: PM10 RDD_S before lock down and during lock down ph-1,2,3,4 and after first unlock

The above map shows that RDD load is higher in dense populated area even in sitting condition except Bawana as it is industrial area, hence RDD load is more despite being non-dense populated area.

The Buffer tool used in ArcGIS to create a new coverage of buffer polygons around specified input coverage features. In our case the feature is PM2.5, PM10 concentration around monitoring station.

Buffer tool has been used to identify or define an area within a specified distance around a feature i.e. 1/2 miles (802 meter) as recommended by United States Environmental Protection Agency and the effect of pollutant is observed on people, who are present after a buffer distance of 802 meter.



With the Zonal Statistics tool, a statistic is calculated for each zone defined by a zone dataset, based on values from another dataset (a value raster). A single output value is computed for every zone in the input zone dataset.

A zone is all the cells in a raster that have the same value, whether or not they are contiguous. The input zone layer defines the shape, values, and locations of the zones. An integer field in the zone input is specified to define the zones. A string field can also be used. Both raster and feature datasets can be used as the zone dataset.

The input value raster contains the input values used in calculating the output statistic for each zone.

Accordingly, above map shows that 250 no. of people per kilometer square are would be affected by toxic release inventory as per the data of CPCB in this study.

3. Results and Discussion

RDD values are calculated for both fine and coarse particles using the data collected from the CPCB portal. Station wise RDD_{2.5} and RDD₁₀ during standing and walking during all phases is shown in Figure 3 to Figure 5. In between 35 stations, it was found that most of the time Bawana has higher RDD_{2.5} and RDD₁₀ values during different phases as it's an industrial area. Sri Aurobindo Marg and ITO, Delhi have low RDD₁₀ values most of the time particularly during lock down

as its dense populated institutional areas and commercial and residential areas respectively. Both the values of RDD_{2.5} and RDD₁₀ decreased during lockdown phase 1, there was a decrease of 58.08% as there was a complete lockdown. After the removal of restriction during different lockdown phases there was an increase in RDD (10 or 2.5) values for about 24.24% in lockdown phase 2 then about 15.12% lockdown phase 3, 11.8% during lockdown phase 4. On the contrary, during unlock phase 1 there decrease in values of RDD(10 or 2.5) of about 21.76% because of the precipitation which suppresses the concentration of ambient particulate matter in the environment [27]. One study was conducted in small city of Portugal in two different phases emergency phase (lockdown phase) and calamity phase (less restricted phase) and its was found a low concentration of PM in the emergency phase ($PM_{2.5} = 5.15 \pm 2.77 \mu\text{g m}^{-3}$; $PM_{10} = 23.30 \pm 21.53 \mu\text{g m}^{-3}$) then in calamity phase ($PM_{2.5} = 30.92 \pm 31.93 \mu\text{g m}^{-3}$; $PM_{10} = 111.27 \pm 104.53 \mu\text{g m}^{-3}$)[34]. Another study was conducted in four cities in Europe and one in China and it found a reduction in PM_{10} concentration of 6% in Nice, 32% in Valencia, 49% in Wuhan, 9 % in Turin and an increase of 2% in Rome while for $PM_{2.5}$ reduction in concentration was 2.9% in Nice, 12.6% in Turin 12.6% in Valencia 36.6% in Wuhan and increase of 10.6% in Rome[36]. For Delhi, before lockdown $PM_{2.5}/PM_{10}$ is 0.555 during lockdown phases its value decreased in lockdown phase 1 $PM_{2.5}/PM_{10}$ ratio was reduce to 0.517 in lockdown phase 2 it reduces to 0.455 but as the restriction remove its value increase up to 0.488 in lockdown phase 3 and 0.504 in lockdown phase 4 again its value decreases in unlock phase 1 0.429 because of the precipitation. The top three RDD values during different lockdown Phases is shown in Table 2

4. Conclusions

The study shows higher Respiratory deposition dose for both $PM_{2.5}$ and PM_{10} in walking and sitting mode before COVID-19 lockdown than during lockdown and unlock conditions. It shows that Delhi is exposed to higher particulate matter during normal days before lockdown and in lockdown phase once RDD values decrease and again by removing the restriction in different lockdown phases RDD value increase due to an increase in construction activity and vehicular traffic. Though after lockdown phase 4 RDD value decreased in unlocking phase 1 due to precipitation which reduce the particulate matter concentration and hence reduce the RDD value. RDD values were higher for 2019 as compared to 2020 because of the restriction during lockdown phases for both pre-monsoon and post-monsoon. Also, RDD walking values were always found to be greater than RDD in the sitting mode which shows that particle deposition is more during physical activity like walking than sitting. Hence RDD is directly related to the concentration of PM and physical activity. This study shows the importance of RDD in determining the health effect due to PM exposure during physical activity like walking and sitting in dense and less dense populated areas and the importance of continued efforts to reduce pollution and implement sustainable transportation and develop policies to maintain lower PM concentration seen during the lockdown. It is also

concluded that results of the empirical study are corroborated perfectly by the GIS study. The population density favours the PM_{2.5} and PM₁₀ RDD_W and RDD_S loads. The study also reveals that 250 nos. of people per square km of area would be 100% affected, however those who are away from buffer distance would be less affected relatively.

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