# STUDY OF NANO-SIZE PARTICLE DYNAMICS IN URBAN ROAD MICROENVIRONMENT IN DELHI By

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**Environmental Engineering Department** 

Submitted

In fulfilment of the requirements of the degree

of

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to



# DEPARTMENT OF ENVIRONMENTAL ENGINEERING DELHI TECHNOLOGICAL UNIVERSITY

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# STUDY OF NANO-SIZE PARTICLE DYNAMICS IN URBAN ROAD MICROENVIRONMENT IN DELHI

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By KANAGARAJ RAJAGOPAL Roll No. 2K19/PhDEN/504

#### ENVIRONMENTAL ENGINEERING DEPARTMENT

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MAY 2025

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# DECLARATION

I hereby declare that the research work presented in this thesis entitled " **Study of nano-size particle dynamics in urban road microenvironment in Delhi**" is original and carried out by me under the supervision of Dr. Rajeev Kumar Mishra, Associate Professor, Department of Environmental Engineering, Delhi Technological University, Delhi, and co-supervision of Prof. S. Ramachandran, Senior Professor, Space and Atmospheric Sciences Division, Physical Research Laboratory, Ahmedabad being submitted for the award of Ph.D. degree to Delhi Technological University, Delhi, India. The content of this thesis has not been submitted either in part or whole to any other university or institute for the award of any degree or diploma.

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## CERTIFICATE

This is to certify that the Ph.D. thesis entitled "**Study of nano-size particle dynamics in urban road microenvironment in Delhi**", is being submitted by Mr. Kanagaraj Rajagopal for the fulfilment of the requirements for the award of the degree of Doctor of Philosophy in Environmental Engineering, to the Department of Environmental Engineering, Delhi Technological University, Delhi, India, is a bonafide record of original research work carried out by him under our supervision. The results embodied in this thesis have not been submitted to any other university or institution for the award of any degree or diploma.

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My Parents

Mr. Rajagopal,

Mrs. Bagavathy

And

My family members & Friends

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#### LIST OF PUBLICATIONS

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19. **Rajagopal**, **K**., S. Ramachandran, Mishra R.K. (Nov 2022). Measurements ofSize Resolved Nanoparticle Concentration and its Distribution in Polluted Urban Environment during Induced Firework Event. Lecture Notes in Civil Engineering 582. https://link.springer.com/book/9789819623587.

#### ABSTRACT

Due to rapid urbanization, Delhi experiences frequent pollution events, and the particulate matter load exceeds the prescribed limit often. This study analyzes nanoparticle (10 to 1090 nm) during different emission scenarios, seasonal and meteorological conditions in two phases: April to June 2021 (Period I) and October to November 2021 (Period II). Period I experienced around 31% less concentration of particles ( $\sim 2.4 \times 10^4$  cm<sup>-3</sup>) due to lockdown restrictions and, on the other hand, particle concentration increased by 35% compared to normal conditions due to the sudden rise in firework emissions in Period II. Except for the post-Diwali phase  $(10^4 \text{ cm}^{-3} \text{ to } 10^5 \text{ cm}^{-3})$ , the concentrations lie between 10<sup>3</sup> cm<sup>-3</sup> and 10<sup>5</sup> cm<sup>-3</sup>. The Aitken modes contribute 10 to 30% of total concentration in both periods. Particles in nucleation and accumulation modes contribute 30 to 40%, 20 to 30%, 15 to 25%, and 35 to 50% in Periods I and II, respectively. Concentration and behavior of nano particles in different seasons (winter, spring, summer, monsoon, and autumn) are examined, for the first time. Concentration of particles is classified into four different sizes as N<sub>nuc</sub> (10 to 30 nm, nucleation), N<sub>satk</sub> (30 to 50 nm, small Aitken), N<sub>latk</sub> (50 to 100 nm, large Aitken), and Nacc (100 to 1000 nm, accumulation mode), and the total (10 to 1000 nm) particle number concentration (PNC) as N<sub>total</sub>. PNC ranges between  $10^4$  cm<sup>-3</sup> and  $10^6$  cm<sup>-3</sup> over Delhi during the year, and the highest concentration occurs in winter. Winter concentration is 2 times higher than monsoon, summer, autumn and spring concentrations, respectively. Nnuc, Nsatk, Nlatk and Nacc and their respective contributions to total PNC exhibit significant seasonal variations. During winter Nlatk and Nacc contribute more to total PNC due to coagulation, with N<sub>acc</sub> alone contributing >40% to total PNC. N<sub>nuc</sub>, N<sub>satk</sub>, and N<sub>latk</sub> are higher in spring and summer during mid-day due to nucleation and/or ultrafine particle burst events. The direct primary emissions from engine exhaust produce a prominent double hump structure during morning and evening peak hours in winter and autumn. PNC and their contributions exhibit day-night variations as they are influenced by emission sources and variations in meteorological parameters (wind speed, relative humidity, temperature, solar radiation and boundary layer height) between day and night. Carbon monoxide correlates positively with Nacc in all seasons ( $R^2 \sim 0.5$ ) as fossil fuel emission is predominant source for gases and particles in study environment. The concentration of UFP size range particles is dominantly higher (70 to 80%) during peak hours than the non-peak hours. The variations in particle number concentration depend on the intensity and emissions of sources during peak, and nonpeak hours. UFP contributes ~60 to 80% to the total particle number concentration in the urban roadside microenvironment, and its contribution increases during peak hours. During the winter season, the average total particle number concentration was observed to be maximum ( $4.1 \times 10^4$  cm<sup>-3</sup>) with a higher surface area of particles of  $3.5 \times 10^{-3}$  mm<sup>2</sup>m<sup>-</sup> <sup>3</sup>. Compared to the monsoon season, the concentration of NO<sub>X</sub> was 5 times higher in winter. The boundary layer height in the study region ranged from 600 to 2400 m during different seasons, and the maximum ventilation coefficient was observed to be >3000 m2s-1 during summer. Precipitation reduced the concentration of particles by half, from 2.2 x  $10^4$  to 1.1 x  $10^4$  cm<sup>-3</sup>, due to wet scavenging. The study revealed that the concentrations of particles depend not only on primary emissions but also are influenced by local meteorology and other co-emitted pollutants. The Multiple Path Particle Dosimetry (MPPD) model simulated values show that the order of deposition goes as alveoli > bronchiole > bronchus. The deposition in the study area ranges between 10 and 18 million nanoparticles during different hours of the day, whereas the estimated inhalable particles vary between 0.5 to 1 billion. The concentration of total inhalable particles and the actual particles deposited in the lung varies. The seasonal sequence of deposition of nanoparticles is winter > monsoon > summer > autumn > spring. The

deposition of nanoparticle in adults is 30 to 40% higher than in children and infants, and further, the deposition is higher in the alveolar region than in the bronchiole and trachea regions. About 90% of the particles get deposited in the alveolar regions, 6 to 8% in the bronchiole region, and 2% in the trachea region. The estimated deposition of nanoparticles for an individual working 8 hours a day in the near road conditions is 338  $\mu$ g/year in Delhi. The deposition increases almost linearly as a function of time, and is 3 times higher (1016  $\mu$ g/year) for a person residing near the road throughout the day (24 h). The deposition fraction of particles ranges between 0.05 and 0.10  $\mu$ g/day in alveolar region,  $<0.05 \mu g/day$  in the bronchiole region, and lies between 0.02 and 0.04  $\mu g/day$  in the trachea. The nanoparticles deposited in the respiratory system can lead to the development of various diseases such as asthma, chronic obstructive pulmonary disease, and can lead to carcinogenicity. Number concentration-based studies are essential for estimating the potential impacts on human health due to air pollution. The study provides information regarding vehicle emission-based particle concentration under various emission scenarios in urban cities, which is crucial for estimation of emissions, health impact assessment, future policy formulation and strategy measures. These quantitative results on seasonal variations of air pollutants together with the knowledge on seasonal variations in meteorological parameters and atmospheric dynamics provide a foundation which can positively contribute to the planning and devising mitigation measures aimed at improving air quality and public health. The study provides new insights on inhalable particle concentration during the day that are crucial for strategy development, emission mitigation, and health hazard assessment for the citizens. Understanding the dynamics of atmospheric nanoparticles in urban roadside environments provides on the deposition of nanoparticles in humans residing near roadside conditions are crucial to estimate the

human health risk potential, and to formulate mitigation measures for exposure reduction which can result in a better and sustainable future.

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#### LIST OF ABBREVIATIONS

- ANOVA Analysis of Variance
- AQI Air Quality Index
- AT Atmospheric Temperature
- BAU Business as Usual
- CCN Cloud Condensation Nuclei
- CFD Computational Fluid Dynamics
- CNG Compressed Natural Gas
- COVID Coronavirus Disease
- CPC Condensation Particle Counter
- CPCB Centre Pollution Control Board
- CPMA Centrifugal Measurement of Particle Mass
- CS Condensation Sink
- DLPI Dekati Low Pressure Impactor
- DMA Differential Mobility Analyzer
- DMPS Differential Mobility Particle Sizers
- DMS Differential Mobility Spectrometer
- DPF Diesel Particulate Filter
- EAA Electrical Aerosol Analyzer
- EC Elementary Carbon

EDA	Exploratory Data Analysis	
ELPI	Electrical Low-Pressure Impactor	
EPA	Environmental Protection Agency	
FIMS	Fast Integrated Mobility Spectrometer	
FMPS	Fast Mobility Particle Sizer	
FR	Formation Rate	
GAM	Generalized Additive Model	
GIS	Geographic Information System	
GMD	Geometric Mean Diameter	
GR	Growth Rate	
HCV	Heavy Commercial Vehicle	
HDV	Heavy-Duty Vehicles	
HEI	Health Effects Institute	
ICP-MS	Inductively Coupled Plasma Mass Spectrometry	
IPCC	Intergovernmental Panel on Climate Change	
KW-Test	Kruskal-Wallis Test	
LCV	Light Commercial Vehicles	
LDMA	Long Differential Mobility Analyzer	
LDV	Light Duty Vehicles	
LII	Laser Induced Incandescence	
LPI	Low Pressure Impactor	

- LUR Land Use Regression MPPD Multiple Path Particle Dosimeter model NAAQS National Ambient Air Quality Standards NCR National Capital Region NOAA National Oceanic and Atmospheric Administration NPF New Particle Formation OPC **Optical Particle Counter** PMC Particle Mass Concentration PNC Particle Number Concentration PND Particle Number Distribution Particle Number Size Distribution PNSD PTEF Polytetrafluoroethylene Membrane Filters RDD **Respirable Depository Dose** RH **Relative Humidity** ROS **Reactive Oxygen Species** SD **Standard Deviation** SEM Scanning Electron Microscope **SLPE State-Level Public Enterprises SMPS** Scanning Mobility Particle Sizer Solar Radiation SR
- TEM Transmission Electron Microscopy

- TEOM Tapered Element Oscillating Microbalances
- UFP Ultra Fine Particulates
- URMe Urban Road Microenvironment
- VOC Volatile Organic Compounds
- WHO World Health Organization
- WRF Weather Research Forecast
- WS Wind Speed

# CHAPTER-1 INTRODUCTION

#### **1.1 Background**

Environmental pollution is a major concern globally (Rentschler and Leonova, 2023). Pollution not only affects the features of the environment but also poses a serious threat to human health and other living creatures (Jbaily et al., 2022). Air is one of the major components of the environment, and on average, around 12kg of air is consumed by an individual for survival (Kulshreshtha and Khare, 2010). Air intake is 12 to 15 times more than the volume of food consumed (Reggente et al., 2015). The pollutants present in the air are more harmful than the pollutants present in the food (De Nazelle et al., 2017). Clean air is free from any impurities such as solid, liquid, or gaseous substance, which is essential for human beings (Patel et al., 2021). Air pollution is becoming a major issue in all major Indian cities (Kumar et al., 2012). Both natural and anthropogenic sources cause the air pollution (Resmi et al., 2019). The natural sources include volcanic eruption, forest fires, pollen, and occasionally lightning activities. Similarly, anthropogenic activities involve industrial processes, burning fossil fuels, automobiles, etc (Moreno-Ríos et al., 2022). The emissions released from natural activities are occasional (Junkermann and Hacker, 2022). Also, they remain in the atmosphere for a short period, whereas in anthropogenic activities, the emissions are continuous throughout the year, and they are emitted into the atmosphere in regular intervals, which also stays for a long time due to the frequent emissions (Lv et al., 2020).

The issue of air pollution in urban regions began in the 20<sup>th</sup> century from various sources that use fossil fuels. The various sources include smoke from factories, furnaces, boilers, ovens, steam engines, exhaust fumes from automobiles, powerplant emissions, major industries such as oil and zinc refineries, chemical and meteorological industries, iron and steel plants, food industries contribute to the different types of air pollution (Zhu et al., 2022). Among all these sources, automobile sources contribute to more urban emissions (Al–Dabbous et al., 2017). Air pollution is broadly classified into primary and secondary pollutants (Table 1.1). Primary pollutants are emitted directly into the atmosphere through identifiable sources, either

by natural or anthropogenic activities. The major primary pollutants include suspended particulate matter, oxides of nitrogen (NO, NO<sub>2</sub>, NO<sub>x</sub>), Sulphur (SO<sub>2</sub>), carbon (CO, CO<sub>2</sub>), and volatile organic compounds (VOCs) (Yadav et al., 2021.). These primary pollutants react with each other or with other compounds in the atmosphere, such as water vapor, with or without the influence of sunlight, to form different types of pollutants, which are referred to as secondary pollutants (Zhang et al., 2021). The secondary pollutants are chemical substances that are produced from the chemical reactions of natural or anthropogenic sources. Major secondary pollutants are sulphuric acid (H<sub>2</sub>SO<sub>4</sub>), ozone (O<sub>3</sub>), formaldehyde, and Peroxy-acryl-nitrate (PAN) (Nelson et al., 2022.).

The suspended particulate matter in the air is found in solid or liquid form in the atmosphere and is based on the size of the particles. The majorly known particulate matter are Particulate matter with a size lesser than  $10\mu m$  (PM<sub>10</sub>) or particles with a size lesser than 2.5  $\mu m$  (PM<sub>2.5</sub>) (Fatima et al., 2022). The particles larger than a molecule but small enough to remain suspended in the air are called aerosols.

Type of particles	Terminology	Meaning	Examples
Liquid particles	Mist	Aerosols, which	Sulphuric acid
		have liquid droplets	mist
	Fog	Aerosols, which	
		have water droplets	
Solid particles	Dust	Solid aerosol	Dust storm
		particles are	
		generated through	
		the grinding down	
		of larger particles.	
	Smoke	Solid aerosol	Cigarette smoke,
		particles produced	smoke from the
		by chemical	burning garbage
		reactions	

Table 1.1 Classification of particulates in suspended matter (Hinds, et al., 1999)

Fumes	Similar to smoke Zinc/lead fumes.
	but are produced by
	condensation of hot
	vapor of metals

The aerosol particle concentration in the ambient atmosphere may range from  $10^6$  to  $10^7$  cm<sup>-3</sup>, irrespective of the urban or remote environment (Baldauf et al., 2013). Highly polluted urban regions can exhibit a higher magnitude of concentration. The diameter of the particles ranges from a few nanometers to  $100 \ \mu m$  (Meier et al., 2015). The aerosol particles from combustion processes such as automobiles, power generation, and wood burning can produce particles from 10 nm to 1  $\mu m$  (Manigrasso et al., 2020). The aerosols generated from windblown dust, pollens, plant fragments, and sea salts are generally larger than 1  $\mu m$  (Garg, 2018). The particles produced in the atmosphere through the photochemical process are typically found to be smaller than 1  $\mu m$  (Gerling and Weber, 2022). The particles' properties and size also vary significantly based on the generation size. The classification of particles of different sizes based on the sources is shown in Fig.1.1.

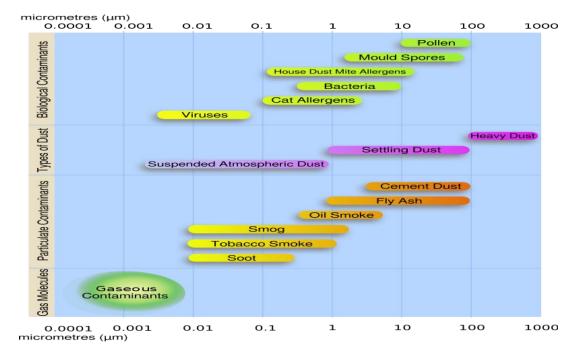


Figure: 1.1 Different types and sizes of major atmospheric particles (Hawk et al., 2020)

Generally,  $PM_{10}$ , and  $PM_{2.5}$  particulate matter, is measured in mass concentration basic, but ultrafine and nanoparticles (due to negligible mass) are estimated based on the particle number per unit volume as particle number per cm<sup>3</sup> (PNC) (Sarangi et al., 2018). The higher-size particles, calculated in mass concentration, show less numbers than the particles smaller particles, or the aerosols are measured in number concentration, which will be found in more numbers (Gerling et al., 2020). However, their mass will be low (Fig.1.2). The aerosol particles generated through various processes are further classified into nucleation (10 to 30 nm), Aitken (30 to 100nm), and accumulation mode (100 to 1000nm) particles based on their size distribution (Agudelo-Castañeda et al., 2013). The Aitken mode particles can be further classified into small Aitken (30 to 50 nm) and large Aitken mode (50 to 100 nm) particles. The emissions from the transportation sector with size ranges (10 to 100 nm) are often called ultrafine particles (Sabaliauskas et al., 2013). The ultrafine particles are ubiquitous in the environment, and this size range of particles can contribute up to 70 to 80% of the total particle number concentration (Zhu et al., 2002).

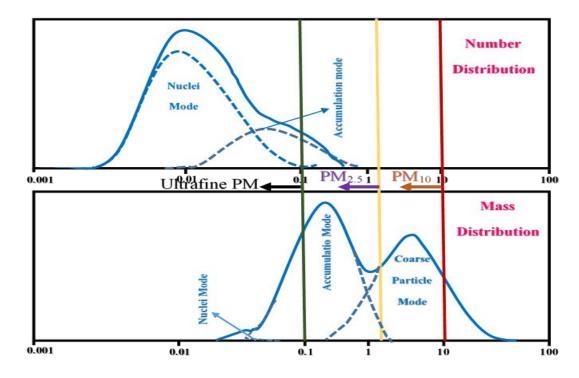


Figure.1.2. UFP size and mass distribution (Adopted from – Hinds 1999)

#### 1.2 Need for the study

Delhi is the national capital of the country and one of the highly urbanized cities where air quality is a major concern. The air quality perspective of India 2023 shows that the daily average concentration of PM<sub>10</sub> is 205  $\mu$ g/m<sup>3</sup>, and for PM<sub>2.5</sub>, it was 100  $\mu$ g/m<sup>3</sup>, which is higher than the prescribed standards (Arub et al., 2021). The cumulative analysis for the years 1990 to 2022 in Delhi shows that the contribution of vehicular sources for air pollution is more, i.e., 10 to 35%, followed by constructional activities (10 to 30 %), industrial activities (10 to 30%), cooking activities (< 10 %) in summer and < 30% in winter, open waste burning contributes to around 5 to 15%, dust storm activities up to 5%, agricultural residue burning up to 3% (Jose et al., 2021). Air quality is directly associated with human health, and the recent report by the Indian state-level disease burden initiative estimates that the death rate due to outdoor air quality has increased by 115 percent (Bhandari et al., 2020). The northern and central parts of the country face more economic losses due to air pollution, around 1.4 % of GDP, and increased premature deaths. In 2019, the country attributed to around 67 million deaths, accounting for around 17 % of the total deaths in the country (Manojkumar and Srimuruganandam, 2021). Among them, outdoor air pollution contributed to around 0.98 million, and indoor or household air pollution contributed to around 0.61 million.

On the other hand, the death rate due to outdoor air quality increased to around 115% (Dahari et al., 2022). Delhi records the highest per capita economic losses due to air pollution, followed by Haryana. Out of eight deaths in India, one is attributed to air pollution as per the Global burden of diseases (Machaczka et al., 2021). Air pollution is one of the risk factors for death in India, so it is necessary to do various research activities to estimate, control, and mitigate air quality in this domain (Hussein et al., 2022). Various studies are conducted in the study domain, estimating various concentrations of particulate matter and gaseous pollutants in its sources. Very few studies have been done to estimate the ambient concentration of aerosols/ultrafine particles. The region receives most of the emissions from the transportation sector, so estimating the aerosols concerning the transport sector is necessary to develop strategies for reducing the emissions from the transportation sector. Another major

need for the study is to estimate the potential human health impact assessment due to these emissions. The particles of lesser size will have more penetration into the human respiratory system and will seriously threaten human health, so quantifying the particles is necessary to estimate the residents' exposure.

#### **1.3 Objectives of the study**

Considering the impact of the transportation sector and its contribution to air pollution, the research has been undertaken with the following research objectives.

1. To measure and quantify the nano-size particles in urban roadside microenvironment (URME).

2. To study the role of meteorological parameters in particle number size distribution (PNSD).

3. To investigate the nano-size particle dynamics over different seasons at a selected transport corridor.

#### 1.4 Organization of the Thesis

The thesis comprises five major chapters, and the brief outline is provided below chapter-wise.

**Chapter 1** deals with the introduction, which consists of different types and sources of air pollution and status of air quality in Delhi (study region) and displays the study's major objectives with a special emphasis on the need for the study. This chapter establishes the platform for writing the subsequent thesis chapters, such as the literature review and result and discussion section.

**Chapter 2** includes a comprehensive literature survey which is available related to the study. The literature review is in different sections, such as the scenario of air pollution in Delhi, how the concentration of air pollution and sources changes during different seasons of Delhi, various studies done globally to estimate the particle number concentration, size distribution of the particles in the different environmental conditions and sources. The literature review also includes dynamics of the variation of particle numbers based on the sources. The chapter also includes the various types

of measurement instruments and the concentration status in different countries. The final section of the chapter includes literature related to health impacts, its penetration into the human body, and its deposition in different regions of the respiratory system.

**Chapter 3** represents the information regarding the study area where the study is conducted, along with the information regarding the instrument used. The methodology adopted to conduct the study, various instruments used for data collection, the software used for data analysis and interpretation, the model used for health impact assessment, and other statistical tools used for the study are discussed in this chapter.

**Chapter 4** includes detailed results and a discussion section in which the first section contains quantification and measurements of these aerosol particles/ nanoparticles in the urban roadside microenvironments of Delhi, along with their seasonal dynamics. The variation of concentration is due to changes in the vehicular density due to the restrictions. The section also estimates the role of meteorological parameters such as relative humidity, wind speed and direction, temperature, solar radiation, and boundary layer conditions, along with the meteorological parameters of the local emissions in terms of gaseous pollutants such as SO<sub>2</sub>, NO<sub>2</sub>, CO, O<sub>3</sub>, and BC is analyzed in this chapter. The final section of the chapter uses mathematical and computational models for analyzing the health impact assessment of the particles in the human respiratory system along with their deposition fraction. The estimated total particle deposited in the different sections of the lungs is provided.

**Chapter 5** describes the detailed conclusion of the experimental studies, its major findings, and a study summary. The results also highlight the effects of the vehicular sources in determining the concentration of the particle's variation.

#### 1.5 Summary

The different chapters of the study will provide detailed discussion about the atmospheric nanoparticle's concentration in the urban roadside microenvironment in the Delhi city.

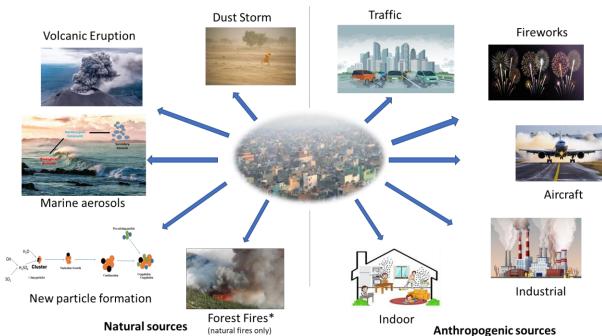
# CHAPTER -2 LITERATURE REVIEW

#### 2.1 Introduction

The atmospheric pollutants differ in their properties and based on their properties the pollutants are classified as particulate matter and gaseous pollutants (Vu et al., 2017). Nanoparticles are a complex mixture of aerosols and small particles suspended in the air with different chemical characteristics (Sly and Schüepp, 2012). The particles size range starts from few nanometres to micrometres (Jayaratne et al., 2015). The nanoparticles are measured and identified using their electrical mobility diameter instead of their chemical composition (Jeong et al., 2021). The nanoparticles are further classified into different size ranges by different researchers and the widely accepted classifications are, Nucleation particles (N<sub>nuc</sub> - 1 to 30nm), Aitken mode particles 30 to 100nm which are further classified into small (N<sub>satk</sub> - 30 to 50nm) and large Aitken mode particles ( $N_{satk} - 50$  to 100nm), accumulation mode particles (100 to 1000nm) (Kumar et al., 2011). The particles in the size range of < 100 nm are also referred as ultrafine particles (UFP) and the particles < 300nm are collectively called as atmospheric nanoparticles (ANP) (Kumar et al., 2012). The particles in the range of 250 to 500nm are referred widely as quasi- ultrafine particles(qUFP) (Yadav et al., 2022a). Current air quality standards don't monitor the concentration of these nanoparticles but these particles influence around 80 to 90% of the total nanoparticle's concentration (Sabaliauskas et al., 2013). However, the nanoparticles contribution for the mass of the particles such as Particulate matter (PM<sub>2.5</sub> and PM<sub>10</sub>) (Mishra et al., 2015).

The nanoparticles are measured in the particle number concentration and are often referred as PNC (Lin et al., 2022). In urban regions the total PNC of the nanoparticles are dominated by the contribution of transportation sources (Zhu et al., 2002). The modal classifications are primarily used in the aerosol research and later adopted for nanoparticles study to classify anthropogenic emissions (Dallosto et al., 2013). The number concentration and the size distribution of the atmospheric nanoparticles varies significantly based on the sources, geography and climatic conditions of the region (Posselt et al., 2019). The size distribution reveals the secondary transformation

process such as formation (Wang et al., 2011). The number concentration and size distribution of the particles in the urban regions varies based on the source's intensity and its interaction in the atmosphere which determines different process in the atmosphere (Sabaliauskas, et al., 2012). Different mode nanoparticles also indicate the different sources as well. In the urban regions the concentration of the nanoparticles was seen higher at their emission sources and its size distribution increase with respect to time and distance from the sources (Hudda et al., 2012). During the natural process such as new particles formation (Kompalli et al., 2014) the nucleation mode particles concentration was found more whereas during the fireworks emission the accumulation mode particles as soot mode particles in the Aitken mode particles. Studies refer accumulation mode particles as soot mode particles. The aged traffic emissions are found to be in the accumulation mode particles (Xie et al., 2021).



Sources of Urban atmospheric nanoparticle emissions

# Fig 2.1 Different sources of urban ultrafine particles

The measurements of these atmospheric nanoparticles are done through highly sophisticated expensive instruments which use total particle number concentration measurements (Machaczka et al., 2021). The upper and lower size range of the detection of these particles' changes from one manufacturer to other manufacture. In general, the condensation particle counter with electrically mobility measurement are the widely used principle for the measurement of these nanoparticles in the atmosphere (Aggarwal et al., 2012). Few instruments use optical measurements for measuring the nanoparticles found in the atmosphere. The scanning mobility particle sizer and differential mobility particle sizers are used by different research throughout the world for the measurements (Hillemann et al., 2014).

# 2.2 Sources of nanoparticles in urban atmosphere

In urban regions the nanoparticles exist due to both natural and anthropogenic sources (Kumar and Gupta, 2013) (Table 2.1). The natural process which contributes for the nanoparticle's emissions are volcanic eruptions, natural forest fires, marine aerosols, dust storm emissions and new particles formation activities (Fig 2.1) (Masiol et al., 2018). Similarly, the anthropogenic activities include traffic emissions, fireworks, aircraft sources, industrial and indoor sources contribute for the primary nanoparticle's emission from the anthropogenic sources (Meskhidze et al., 2019). The secondary sources of nanoparticles in the atmosphere are due to the condensation and coagulation of the primarily emitted nanoparticles (Saha et al., 2019). The new particle formation events, coagulation and condensation of the nanoparticles are influenced by the atmospheric conditions where the emissions took place and then based on that the frequency and efficiency of the process differs. In general, urban regions nanoparticles concentration are dominated by the emission from the transportation sectors that includes different vehicles such as two, three wheelers, light and heavy motor vehicles (Kumar et al., 2014). The automobiles include aircraft sources and marine traffic sources as well (Zhao et al., 2021).

Several studies revealed that the nanoparticles concentration of ultrafine particles in the total nanoparticle's concentration is around 70 to 95 % due to the transportation sources. Globally several studies about the nanoparticles concentration in the atmosphere is conducted in the developed nations such as USA (Saha et al., 2019), Germany (Giemsa et al., 2021; J. Sun et al., 2020), Europe (de Nazelle et al., 2017;

Garcia-Marlès et al., 2024) and few studies are found in the developing nations such as India (Kompalli et al., 2016; Yadav et al., 2021), China (Yin et al., 2019; Zhou et al., 2020; Zhu et al., 2002), Japan (Orikasa et al., 2020), and other few south Asian countries (Akteruzzaman et al., 2023). Very few studies are conducted in the developing nations. The European union developed the emission norms for the nanoparticles with a standard of  $6 \times 10^{11}$ /km for the gasoline and diesel engine followed that the World Health Organisation (WHO) emphasized the importance of importance of atmospheric nanoparticles and came up with a guideline for the ambient nanoparticle's concentration as low and high category. The low with a concentration of < 1000 particles/ cm<sup>-3</sup> (24 hr mean) and high concentrations with (> 10,000 particles/ cm<sup>-3</sup> for 24 hrs and > 20,000 particles/ cm<sup>-3</sup> for 1 hr) for having particles size ≥10 nm with no upper limit. The WHO in the year 2021 also stated that the measurements of these nanoparticles are to be done with the existing monitoring systems and also the epidemiological studies should be initiated to for the health impacts assessment.

S.no	Country, city	Environment	Concentration (cm-
			3 ×10 <sup>3</sup> )
1.	Australia	Urban Region	26
2.	Bangladesh	Urban background	25
		Urban Region	60
		Urban Roadside	82
3.	Belgium	Urban Region	13
4.	Canada	Urban Region	26
5.	China	Urban Region	33
6.	Denmark	Urban Background	18
7.	Europe	Urban Background	8
		Urban Roadside	19
		Urban Region	17
8.	Finland	Urban Region	35
		Urban Roadside	40

Table 2.1 Summary of atmospheric nanoparticles around the world

9.	France	Urban Background	8
		Urban Roadside	27
10.	Germany	Urban Background	17
		Urban Roadside	60
		Urban Region	31
11.	India	Urban Background	12
		Urban Roadside	45
		Urban Region	27
12.	Italy	Urban Region	45
13.	Jordon	Urban Region	35
14	Malaysia	Urban Region	153
15.	Netherlands	Urban Region	35
16.	Singapore	Urban Region	12
17.	Sweden	Urban Roadside	65
18.	Taiwan	Urban Roadside	120
19.	United Kingdom	Urban Background	7
		Urban Roadside	71
		Urban Region	17
20.	United states of America	Urban Background	8
		Urban Roadside	15
		Urban Roadside	30
		Urban Roadside	22

# 2.3 Urban roadside microenvironment nanoparticles

In urban regions, roadside microenvironments are hubs for the nanoparticle's emission (Zhu et al., 2004). The concentration observed in the roadside is combination of both vehicular and ambient background emissions in the region (Zhu et al., 2004). The concentration in the roadside microenvironment can be as high as 25 times compared to the urban background conditions. The traffic density is another one deciding factor for the nanoparticle's concentration. The developing countries vehicle population is

increasing at an alarming rate, especially in Delhi the vehicle population will increase from 4.74 million in 2010 to 25.6 million in 2030 as projected. The traffic intersections such as red signal region revive more concentration of about 3.5 times higher than the green signal intersections (Y. Wang et al., 2008). The roadside microenvironment has significant impact on human health due to the location of residential areas adjacent to the roadside microenvironments in urban regions (Hagler et al., 2010; Sabaliauskas, Jeong, et al., 2012). The concentration of the nanoparticles in the urban street canyon is much higher than the urban roadside microenvironment due to the vortex effect (Kumar et al., 2008a). Majority of the global urban cities experience this canyon effect due to their built environment in urban regions (Weichenthal et al., 2014). The concentration of the nanoparticles in the urban microenvironment also depends on the driving pattern (Goel and Kumar, 2014). The nanoparticle emission is found to be high during low speed conditions and waiting period. The waiting period concentration increases around 14 times compared to the normal condition and also the accelerating conditions are found to have 6 times higher concentration of nanoparticles compared to the idling conditions (Argyropoulos et al., 2016).

The nanoparticles concentration in the roadside environment also changes during the different hours of the day such as peak and non-peak hours (Farrell et al., 2016). The particle number concentration of the nanoparticles is found similar to the traffic pattern and the size distribution exhibit bimodal distribution of the particles (Agudelo-Castañeda et al., 2013). The concentration of the nanoparticles in the tailpipe can reach up to  $10^7$  to  $10^9$  cm<sup>-1</sup> whereas, the new particle formation can add particles up to  $10^3$  to  $10^4$  cm<sup>-1</sup> under high condensation sink and growth rate so in urban region the vehicular sources dominate the total particle number concentration (Goel and Kumar, 2016). Asian countries report the higher concentration of urban nanoparticles compared to the Europe. with a mean concentration of about as  $6.65 \pm 2.88 \times 10^4$  cm<sup>-3</sup>,  $4.81 \pm 2.61 \times 10^4$  cm<sup>-3</sup> and  $3.78 \pm 2.01 \times 10^4$  cm<sup>-3</sup> (Nabizadeh et al., 2018). Congested traffic, a common phenomenon in urban cities can increase the nanoparticles concentration in roadside microenvironments when the vehicle constantly accelerating and deaccelerating (Joerger and Pryor, 2018). The urban regions in developing nations experience soaring levels of atmospheric nanoparticles but lack of scientific studies

and un availability of data base makes it hard to provide necessary mitigation measures (Belkacem et al., 2020). The nanoparticles emissions vary based on vehicle types, fuel quality, different driving patterns and in cabin measurements are still a grey area which needs to be focused (Belkacem et al., 2022).

The background concentration in the roadside environment also varies based on the location, surrounding industrial emission which also influences the concentration in the roadside environment (Zheng et al., 2022). Studies showed that the exposure of nanoparticles in the vicinity of road is a magnitude higher that the ambient levels (Yao et al., 2022). The exposure occurs during transportation, waiting for transit, walking, residing in homes near road environments, working in vicinity of road (Rajagopal et al., 2023).

# 2.4 Different urban microenvironments

In urban regions apart from roadside microenvironments, tunnel environments are another important region where the concentration of the nanoparticles increases significantly (Y. H. Cheng et al., 2010). A study conducted in Taiwan in a tunnel environment shows that the concentration of the nanoparticles inside the tunnel were 19 to 70 times higher than the roadside environment (S. C. Chen et al., 2010; Cheung et al., 2016). When the vehicle fleet inside the tunnel increases from 100 to 2000 then the concentration of the nanoparticles also increases from  $49 \times 10^3$  cm<sup>-3</sup> to  $125 \times 10^3$  cm<sup>-</sup> <sup>3</sup> (Kearney et al., 2011). The study measures the nanoparticles concentration during different vehicle density conditions and the found that the traffic volume has a significant impact on nanoparticles concentration and also the concentrations are imprisoned inside the tunnel due to the confined space (Zhao et al., 2020). Similarly, another study found the concentration in the peak hours and how the fresh emissions vary its size modal distribution inside the tunnel due to the atmospheric activity with respect to the time (Skuland et al., 2022). The size distribution study reveals that outside the tunnel nucleation mode particles contributes more due to the fresh emissions and inside the tunnel Aitken mode particles due to the coagulation of particles (Skuland et al., 2022).

Urban residential areas are another microenvironment where more studies need to be conducted (Chen et al., 2020). Preliminary study by researchers found that the concentration in the urban residential areas are high during the winter periods with a concentration of about 10  $\times 10^3$  cm<sup>-3</sup> and less during summer 7  $\times 10^3$  cm<sup>-3</sup>. In the residential regions the concentration of 10 to 100 nm particles is found more than 100 to 800 nm particles and the influence of outdoor concentrations are also found more in the residential zones (Hussein et al., 2019). Urban street canyons are another major microenvironment in the urban regions where the higher levels of atmospheric nanoparticles are found (Weber et al., 2013). The concentration increases due to the recirculation and vortex formation of the exhaust emission due to the infrastructure (Gerling and Weber, 2022). The study in different countries reveals that urban street canyon concentration in Spain is 13×10<sup>3</sup> cm<sup>-3</sup>, 17×10<sup>3</sup> cm<sup>-3</sup> in Germany, 40×10<sup>3</sup> cm<sup>-3</sup> in Finland and 60 to 70×10<sup>3</sup> cm<sup>-3</sup> in Sweden. Metro corridors are another major microenvironment in the urban regions where large number of people's commute (Yang et al., 2021). In Barcelona the metro station study revealed a concentration of about 23×10<sup>3</sup> cm<sup>-3</sup> similarly, in USA- 25×10<sup>3</sup> cm<sup>-3</sup>, Helsinki - 31×10<sup>3</sup> cm<sup>-3</sup>, 10 to  $29 \times 10^3$  cm<sup>-3</sup> in UK,  $11 \times 10^3$  cm<sup>-3</sup> in chez Republic and  $12 \times 10^3$  cm<sup>-3</sup> in Spain are the globally reported concentrations in the metro stations (H. Patel et al., 2023). The literature suggests that in the urban regions there are different microenvironments where the concentration of the nanoparticles is extremely high compared to the ambient concentration (Bergmann et al., 2022).

# 2.5 Nanoparticles concentration in airport regions

Aircraft sources are one among the major source of atmospheric nanoparticles (Table 2.2) apart from vehicular sources (Ren et al., 2016). The concentration of nanoparticles is found high during the time of take-off and landing in the airport region (Pirhadi et al., 2020). The concentrations in the airport are highly influenced by the wind parameters. The concentration of the nanoparticles in the airport also varies based on the distance of measurement away from source (runway) and also in the UP/Down wind condition (Pirhadi et al., 2020). Few global studies are conducted to measure the concentration of these nanoparticles in the airports (Stafoggia et al., 2016). A study conducted in Tianjin international airport china reports that the concentration of

particles peaks between 25 to 50 nm during the take-off and landing of aircrafts and also the concentration also increases to  $4 \times 10^5$  cm-3 whereas during normal period it was  $2.2 \times 10^5$  cm<sup>-3</sup>. During take-off the concentration increase was ~  $25 \times 10^3$  where as during the landing the increase in concentration was  $\sim 5 \times 10^3$ . The geometric mean diameter of the aircraft nanoparticles is found lesser than the diameter of the particles found in the roadside microenvironments. The aging process of the aircraft emitted nanoparticles are less compared to the other vehicular sources due to their high dilution rate (Yadav, Mishra, et al., 2019). The airport concentrations are influenced by different events in airport such as take-off, landing, ground operations, background existing concentrations and meteorological conditions(Chen et al., 2010). The airport regions receive a mean of ~20% of the concentration contribution from the road traffic sources (Ragettli et al., 2014). The lubrication oil used in Jet engines also emit particles in the range of 10 to 30nm. The percentage contribution of sub10nm particles in the total particle number concentration is high compared to the other urban regions. Heal impacts due to the exposure of these nanoparticles are significant even if their exposure time is low (Moreno-Ríos et al., 2022). Health subjects exposed to atmospheric nanoparticles are found to have change in their forced vital capacity reported in the study conducted by the researchers in the airport environments (Junkermann and Hacker, 2022). The studies also found that the concentrations are time resolved. The different concentration of the atmospheric nanoparticles measured in the different international airports are reported in the table given below.

S.no	Airport name	Country	Concentration (cm <sup>-3</sup> )	Author
1.	Tianjin	China	$22 \times 10^4$	(Ren et al., 2016)
2.	Ciampino	Rome	$1.9  imes 10^4$	(Stafoggia et al., 2016)
3.	Los angles, Hartsfield- Jackson	USA	$1.9  imes 10^4,  0.7  imes 10^4$	(Riley et al., 2016)
4.	Schiphol	Netherlands	$5.3  imes 10^4$	(Lammers et al., 2020)
5.	Narita	Japan	$3.4 \times 10^{17}$ (kg per fuel)	(Takegawa et al., 2021)

Table 2.2 Summary of atmospheric nanoparticles in different airports

6.	Gatwick	London	$9.4  imes 10^4$	(Tremper et al., 2022)
7.	Frankfurt	Germany	$2.4  imes 10^4$	(Dröge et al., 2024)

# 2.6 Role of new particle formation events in urban regions

New particle formation events are observed in different regions such as urban regions (Fig 2.2), rural, free tropospheric, forest areas and coastal zones (Kulmala et al., 2021). In urban regions the new particle formation events usually occur during the late morning and grow throughout the day (Wu et al., 2024). The growth rate of the nanoparticles due to the new particle formation events ranges from 1 to 20nm/hr (Zimmerman et al., 2020). The new particle formation events are occurs based on four main mechanisms (Dinoi et al., 2023). The mechanism includes homogenous binary nucleation, heterogeneous binary, homogeneous ternary nucleation, ion-induced nucleation of binary, ternary based on the environmental conditions (Cheung et al., 2010). New particle formation events are initiated with the help of photochemical reactions and the activity is enhanced by the gaseous precursors such as H<sub>2</sub>SO<sub>4</sub>, NH<sub>3</sub> and other volatile organic compounds (VOC) (Dinoi et al., 2021).

The low volatile vapours in the atmosphere nucleates into neutral molecular clusters which are stabilized by amines, ammonia and organic vapours that are activated by condensation of the low VOCs (Tanda et al., 2019). The factors such as low relative humidity, presence of SO<sub>2</sub> and lesser pre-existing particle concentration, surface area of particles, high insolation and wind speed of the region (Kerminen et al., 2018). The NPF events increases the concentration of nucleation mode particles and it grows subsequently for other modes under favourable conditions in the urban regions (Bousiotis et al., 2021). Urban NPF events can interact with the regional NPF events under certain conditions and grow horizontally in the city regions (Das et al., 2021; C. Deng et al., 2020; Kanawade et al., 2022). The nanoparticles in the urban regions are highly affected by the temperature of the emission sources such as household activities, domestic heating, power production and industrial process and vehicular road emissions flume temperature (Carnerero et al., 2018; Rosati et al., 2021). The cloud condensation nuclei can initiate the NPF events and growth of particles by 40 to 50%

which relates the process of climatic systems which is of global importance (Cheung and Chou, 2013; Zhang et al., 2017).

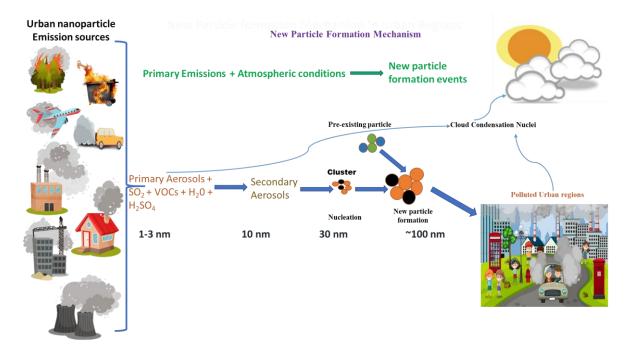


Fig 2.2 New particle formation mechanism in the atmosphere (Yadav et al., 2021)

The global tropospheric aerosol particles are responsible for their contribution to air quality and climate change patterns emphasis the necessity for understanding the particles in a detailed manner including their transformation in the atmosphere (Okuljar et al., 2021). New particle formation events also undergo transformation in size causing increase in concentration of the atmospheric particles in the atmosphere (Wang and Oliver Gao, 2011). Over the years the understanding of the NPF events are evolved due to various breakthroughs in the field and the theoretical prediction of molecular clusters through the Kohler type process (Yu et al., 2020). Previous studies are detected the NPF events in the rural and background regions but the recent several studies in urban regions proved the NPF events in a highly polluted urban region (B. Sarangi et al., 2018). This is due to the low condensations sinks exists in the cleaner regions with high condensation sinks affects the frequency of the NPF events (Yadav et al., 2021). The nucleation of gaseous precursors as nanoparticles in the atmosphere are one of the sources for urban nanoparticle concentration (Wu et al., 2021).

Several studies recommend to investigate the NPF events through micrometeorological perspective including the turbulent fluctuations (Ou et al., 2021). It is estimated that the global tropospheric cloud condensation nuclei are formed through NPF events (30 to 50%) (Chen et al., 2022). Because NPF events are globally phenomenon and it has huge impact on global climate (Blanco-Alegre et al., 2022). The literature suggests that there is a need of further research that includes filed observations that includes different microenvironments of urban regions and different other aspects such as frequency, strength and related meteorology also has to be considered for the study (Rajagopal et al., 2023, 2024). The basic nucleation starts clustering at sub 3nm particles which is also important parameter for measurement in the urban environments for detecting the NPF events and their role in ambient nanoparticles concentration in the urban regions (Jose et al., 2021).

#### 2.7 Nanoparticles emission from fireworks

Firework emissions are considered as the short-term air quality degradation events (Garg, 2018). The most unusual anthropogenic activities that create notable short-term air pollution events are the firework events for celebrating events and festivals throughout the world (Tanda et al., 2019; Zhang et al., 2010). Firecrackers are used to celebrate certain festivals like new year eve, Diwali, spring festivals, Guy fawkes night were extensive fireworks were used for the celebrations (Hoyos et al., 2020; M. Zhang et al., 2010). Literatures reports emission of trace metals, suspended particles, organic compounds, sulphur dioxide and nitrogen oxides but the recent studies revealed that the nanoparticles are also emitted during the firework event and the impact of these emission lasts for few days causing series threat to human beings (Yerramsetti et al., 2013). The studies measured the temporal concentration of the particle concentration which always showed a peak concentration during and after the firework events (Yadav et al., 2022b). Studies conducted during the Chinese New Year firework events concentration of the nanoparticles attains a peak concentration of  $3.8 \times 10^4$  cm<sup>-3</sup> and also the particles in the size range of 100 to 500 nm are found to have more contribution in the total nano particle's concentration (Yao et al., 2022; Yu et al., 2017; Thakur et al., 2010). The Indian subcontinent where Diwali festival is celebrated throughout the country and many studies are conducted during this period for the measurement of the nanoparticles emitted during this period (Chhabra et al., 2020). In the year 2012 the measurements of nanoparticles during Diwali period was reported with a concentration of about  $1.2 \times 10^5$  cm<sup>-3</sup> particles (Saxena et al., 2020). Similarly, several studies were conducted during the different phases of the Diwali where the normal crackers were used then green crackers were used for emission reduction, restriction of crackers bursting and finally banning of crackers bursting in the country (Rajagopal et al., 2024). Series of studies conducted to explain the different concentration of the nanoparticles in the national capital of the India (Yadav, et al., 2019). Studies revealed that during the Diwali the concentration of emission increases by 139 percent comparted to the previous period with a concentration of about  $1.7 \times 10^5$  cm<sup>-3</sup> particles (Prabhu et al., 2019). During the Diwali emissions the shift in the particle's emissions are clearly found in the GMD of the particles which shows a peak of around 180 nm during the fireworks emissions (Vaghmaria et al., 2018). While comparing the green crackers with the traditional crackers the emission of smaller size particles is more compared to the traditional crackers (Izhar et al., 2018; Yadav et al., 2022c). The UFP/N<sub>total</sub> ratio of green crackers are 0.7 where the traditional crackers ration lies between 0.45 to 0.6. Delhi being one of the highly polluted urban regions during Diwali period the concentration of the nanoparticles reaches to a peak of  $2.7 \times 10^5$  cm<sup>-</sup> <sup>3</sup>. Studies that even tough the emission are for short period but the sudden spike in concentration during the Diwali period causes severe health impacts in short and long term (Ghei and Sane, 2018; Sateesh et al., 2018). The emission increases the concentration of inhalable particle concentration in the atmosphere and they may reach up to 10 million particles/day. Whereas the inhalable particle concentration varies based on the different inhalation rates. The concentration may further increase when the individual does any heavy physical activity. The results reveal that the nanoparticles in the atmosphere are causing different illness irrespective of the short term and long-term emissions.

# 2.8 Nanoparticles emission from Non-Exhaust sources

Several studies revealed that the vehicular sources especially exhaust sources are vital sources for the atmospheric nanoparticle's sources in the urban regions (Kumar et al., 2013). But there are several other non-exhaust sources which are also acts as a

mandatory source in the urban regions. The major non exhaust sources for nanoparticle emissions are construction and demolition sources, forest fires/waste burning, powerplants, cigarette smoking, cooking, break and tyre wear and long-range transportation sources are also some of the other major sources in urban regions where the nanoparticles are emitted into the atmosphere (Vu et al., 2015).

# 2.8.1 Construction and demolition sources

The construction sites were the crushing, drilling, usage of machineries, soil excavation, usage of fine sand and cement are the major sources which induce the nanoparticles concentration in the atmosphere (Zhu et al., 2022). Similarly, during the demolition activities usage of different cutters and demolishing activities produces nano emissions. The ambient measurements of nanoparticles during demolition activities reports a concentration 1.6 times higher than the normal back ground concentration (Hopke et al., 2022). Similarly, other studies (Li et al., 2023; Ridolfo et al., 2024) are conducted in a controlled environment for measuring the nanoparticles emissions and the net emissions received during the slab demolition is around  $1.8 \times 10^4$ cm<sup>-3</sup>. The studies also revealed that the nanoparticles emitted during the different demolition process are in the range below 100nm i.e in the range of ultrafine particles (Li et al., 2023). The nanoparticles emitted from the construction and demolition activities are non-volatile in nature whereas the transportation sources emits particles of volatile origin (Ridolfo et al., 2024). The non-volatile nature of the nanoparticles is capable of having longer atmospheric lifetime which is even more vulnerable due to the dispersion in the receptor regions (Teknologi et al., 2016). However, the concentration fluctuates based on the parameters such as the distance of the sampling location from the source.

# 2.8.2 Forest fires and agricultural waste burning

Forest fires and agricultural waste burning emits significant number of nanoparticles in the atmosphere (Lv et al., 2020). The emitted nanoparticles not only cause health impacts but also alters the radiation of the atmosphere due to their property to act as a cloud condensation nuclei (Banerjee and Christian, 2018). In recent years the frequency of the forest fire events increases and it emits fine particles, trace gases and

the emission from the forest fires causes effects in local and regional boundaries due to their transboundary nature (Bhardawaj et al., 2017). Biomass burning for cooking and surface heating also emits nanoparticles into the atmosphere (Marval and Tronville, 2022). The fresh smoke emissions from the biomass burning emits particles in the size range of 100 to 160 nm with a peak concentration at 130nm (Phairuang et al., 2021). Similarly, the aged forest fire smoke emits particles are detected at 250 to 300 nm size ranges (Li et al., 2023). Majority of the studies developed emission factors for the different types of biomass fuels such as savanna fire plumes emit  $3.4 \times 10^{15}$  kg<sup>-</sup> <sup>1</sup> and tropical forest fires emits  $1.5 \times 10^{15}$  kg<sup>-1</sup> (Audignon-Durand et al., 2023). Buring of agricultural residue also emits nanoparticles and their concentration and size distribution also varies from other sources (Li et al., 2017). Studies are conducted for measuring the particle concentration of different fuels under different emissions conditions (Reche et al., 2017). Rice straw burning shows a unimodal distribution of particle emissions in all the three different modes of burning such as open fire, flaming and smouldering (Zhang et al., 2016). Even though the concentration shows a unimodal distribution but the mean diameter of the particle's changes from 52 nm to 141 nm based on the emission mode (Cusack et al., 2013). Wheat straw also shows a unimodal distribution of particles ranging between 10 to 400nm (Dahari et al., 2022). Flaming produces particles in the range of 200 nm where the smouldering produces particles in 500 nm range (Resmi et al., 2019). The studies also show that the nanoparticles emission from the forest fires and agricultural waste products are not uniform and it changes based on types of measurements, type of fuel used and type of combustion involved as well (Donateo et al., 2021).

# 2.8.3 Nano emissions from power plants

Powerplants are well known sources for emission of different types of gaseous and particulate matter (Gerling et al., 2020). In similar fashion the powerplants are emitting nanoparticles as well into the atmosphere (Michaelis et al., 2021). Power plants uses different types of fuel, different ratios of fuel quality which emits different types of nanoparticles (Liu and Cui, 2014). Research studies shows that the nanoparticles emission from the coal fired power plants are multimodal in nature and are emitting particles in the range of  $10^8$  to  $10^{10}$  range before flue gas treatment (Saha et al., 2021).

After flue gas treatment the concentrations reaches to  $10^5$  which is still a higher concentration compared to the background conditions concentrations  $10^3$  (Shrestha et al., 2016). In china a full scaled coal fired power plants were chose for the study and found that the concentration of emission which was  $6.8 \times 10^8$  cm<sup>-3</sup> which 2 order magnitude higher concentration than the higher background PNC concentration  $3.1 \times 10^6$  cm<sup>-3</sup> (Borsós et al., 2012). The electrostatic precipitator used in the cola fired power plants are capable of removing nanoparticles emission concentration from  $10^8$  to  $10^5$  due to their higher performance efficiency of PM removal (Hussein et al., 2022). Cola combustor plant indicates two or three modal concentration mainly sub 100 nm indicates that majority of the particles emitted are in the UFP range (Davulienè et al., 2022). In pre-treated flue gas the bimodal peaks are observed at 9.31 nm and 60 nm. Similarly, peak between 40 to 50 nm, for sulphur bituminous coal and ~80 nm for subbituminous coal are found in different studies(Liu et al., 2021).

# 2.8.4 Nanoparticle emission from cigarette smoking

Tobacco smoking emits different types of toxic substances and pollutants which causes serious health effects for active and passive smokers (Mosonik et al., 2019). In Italy the concentration of the nanoparticles reduced from 7.7  $\times 10^4$  cm<sup>-3</sup> to 3.8  $\times 10^4$  cm<sup>-3</sup> when smoking was banned in a public building (Wu et al., 2021). Water pipe smoking emits particle numbers 5 times higher than the normal cigarette (Fawzy et al., 2024). Detailed study reveals that one cigarette can typically produce particles of about  $\sim 10^{11}$ to  $10^{12}$  cm<sup>-3</sup> particles min<sup>-1</sup> or ~ $10^{12}$  cm<sup>-3</sup> particles cigarette<sup>-1</sup> (Schraufnagel, 2020). The mean concentration of particle number emitted during the smoking of five different brands of cigarette were 3.36  $\pm 0.34 \times 10^{11}$  cm<sup>-3</sup> and similar studies also reports the concentration of about 3.8  $\times 10^{11}$  cm<sup>-3</sup>. Total particles emitted by a single cigarette concentration ranges from 0.64  $\pm 0.19 \times 10^{12}$  particles cigarette<sup>-1</sup>. Few studies report elevated concentration of nanoparticles in the indoor environment which is  $2.7 \times 10^4$ cm<sup>-3</sup> which is 1.5 times higher than the background concentrations. Another study (Dijk et al., 2011) reports elevated concentration of  $2 \times 10^5$  cm<sup>-3</sup> to  $3.5 \times 10^6$  cm<sup>-3</sup>. Studies also found that the peak concentration attained during smoking lasts for 20 minutes and it takes 300 minutes to attain the original background concentration (R. Fuller et al., 2022). The results suggest that the smoking nanoparticles emission not only releases different types of toxic substance into the atmosphere but also, they emit higher concentration of atmospheric nanoparticles which are very harmful especially in the indoor environments.

#### 2.8.5 Nanoparticle emissions from cooking

Cooking activities are found in residential areas, commercial restaurants and open cooking(Xiang et al., 2021). Cooking activities with solid fuels and biomass are know from emission of different types of pollutants but the gas fuels and electric stoves are used widely globally and their emissions are known less (Tang and Pfrang, 2023). During cooking process, the nanoparticles are formed due to the incomplete combustion of food, oil and fuel. Studies revealed that during cooking process the concentration of the PNC increases around 1 to 30 times compared to the background which influences the indoor PNC concentration by 15 times (Pokhrel et al., 2015). During cooking process such as steaming, frying, boiling and type of meal cooked such as vegetables, rice and meat also influences the particle number concentration. Stir frying increases particle concentration about 85 times higher than the normal concentration (Lenz et al., 2023). During cooking process, the concentration increases from  $9.1 \times 10^3$  cm<sup>-3</sup> to  $7.7 \times 10^5$  cm<sup>-3</sup>. Frying of chicken increase particle concentration from 3.7×10<sup>3</sup> cm<sup>-3</sup> to 3.6×10<sup>5</sup> cm<sup>-3</sup> rapidly (Sun and Singer, 2023). The concentration of the nanoparticles changes also based on the different cooking styles such as Indian, Italian, Chinese and American cooking activities (Kang et al., 2023). The stir-frying method is considered to be the most nanoparticle emission cooking methods (Zhang et al., 2021). During cooking hours, the concentration in households increases by 10fold times in living room and 20 to 40 times in kitchen (Kuye and Kumar, 2023). Summary of different cooking related studies shows the concentration of the cooking activities varies between a concentration of  $1.5 \times 10^6$  cm<sup>-3</sup> to  $5.6 \times 10^6$  cm<sup>-3</sup> concentration of particles (Gabdrashova et al., 2021). The emission rates during cooking varies between  $0.35 \times 10^{11}$  min<sup>-1</sup> and  $7.3 \times 10^{11}$  min<sup>-1</sup>. Emission rates for stoving activity is  $51.4 \times 10^{11}$  min<sup>-1</sup> and for toasting it was  $16.7 \times 10^{11}$  min<sup>-1</sup>. During cooking activity, the nanoparticles emissions are varies based on the type of fuel used, food type, cooking method and type of cooking stoves (Delapena et al., 2018). Gas based cooking and electric stove both produce higher number of nanoparticles at high temperature cooking (Gould et al., 2020). Compared to frying technique, boiling way of cooking was found to emit lesser particles. All the different types of cooking emit particles in the size range of particles < 100nm (Kota et al., 2022). The cooking activity in urban regions are also one of the major sources of urban nanoparticles concentration from the non-exhaust sources which has influence on both indoor and outdoor nanoparticles concentration (Zhang et al., 2022).

# 2.8.6 Break wear and tyre wear nanoparticles emissions

Tyre wear and break wear are important non exhaust sources of emissions from the transportation sector (Kumar et al., 2013). In urban regions the concentration of these emission contributes significant amount of emissions to the total emission. The nanoparticles emission from the tyre wear is based on the speed of the vehicle, type of tyre and road condition. In laboratory condition at a fixed speed of 50km/hr and 70 km/hr the nanoparticles emission ranges from 4-30×10<sup>11</sup> km<sup>-1</sup> for studded and nonstudded tyres. The break wear emission emits heavy metal particles such as Fe, Cu, Pb and Zn nanoparticles which are more toxic in nature. The tyre emissions emit highly carcinogenic volatile compounds such as PAHs (Harrison et al., 2018). More break wear emissions are observed in corners, traffic junctions and pedestrian crossing zones. Studies shows that the contribution of wears to the particulate matter is around 16 to 55% in the urban regions and in highways the contribution is  $\sim$ 3%(Kwak et al., 2014). Similar phenomenon will be applicable for the nano emissions from the wear because in urban regions frequent breaking is used compared to the highways. Recent study estimated the payment tyre interface concentration was  $2.5 \times 10^4$  cm<sup>-3</sup> which is ten times higher than the background concentration  $(1-2\times10^3 \text{ cm}^{-3})$ . The mean particle emission ranges from 15 to 50nm which is similar to the exhaust emissions of light motor vehicles. The size distribution showed a peak concentration between 20 to 50nm. The exhaust emission is reducing in urban regions due to change in combustion methods, increasing fuel quality and EV vehicles whereas for non-exhaust the wear emissions are unregulated and keep on increasing due to the increase in vehicle density and congestion.

# 2.9 Urban Indoor Nanoparticles concentration

Nanoparticles emissions in indoor is more vulnerable than the ambient nanoparticles concentration (Nazaroff, 2023). Majority of the time is spent in indoors compared to the outdoors. The major indoor sources include cooking, smoking, hair dryers, use of incense sticks and photo copiers (Kulshreshtha et al., 2008). The detailed summary of cooking and smoking are mentioned in the above sections and in this section other indoor sources are discussed (Okam et al., 2024). In indoor cleaning activity also contributes for indoor nanoparticles exposure (Vicente et al., 2024). After smoking and cooking usage of air fresheners were considered to be the highest particle number concentration contributor (Datta et al., 2017). Burning of candles increases a concentration of particles 10<sup>6</sup> cm<sup>-3</sup> (Zou et al., 2024). Cleaning produces concentration of about 10<sup>5</sup> cm<sup>-3</sup> (Jeong et al., 2023). Steam ironing activity emits more particles compared to ironing without steam (Kulshreshtha and Khare, 2010). In Indoor heating activity was the least nanoparticle emitter in indoor sources (Chen et al., 2020). Fire places have a mean concentration of about 68.5×10<sup>3</sup> cm<sup>-3</sup> during closed fires and  $128 \times 10^3$  cm<sup>-3</sup> during open fires (Pipal et al., 2021). In case of electric heaters different peaks were observed for different manufactures and among them higher concentration was observed for air blow heaters and lower particle concentration was for oil filled heaters (Madureira et al., 2020). Usage of hair dryers generates concentration of particles around 5.3  $\times$  10<sup>4</sup> to 2.5  $\times$ 10<sup>5</sup> cm<sup>-3</sup> with a particle peak attaining between 10 to 25 nm (Kulshreshtha and Khare, 2011). Printing activity in indoor emits mean particle concentration of about 10<sup>4</sup> cm<sup>-3</sup> for 3D printers and for laser printing in indoors for more than 10 pages results in emission concentration of 0.6 to 1.9×10<sup>5</sup> cm<sup>-3</sup> particles and the concentration decreases when measured at a distance more than 2m from the printers. The domestic activities in indoors also emits different types and concentration of nanoparticles and their size distribution also varies based on their source.

### 2.10 Role of vegetative barriers on nanoparticle concentration

The fate of nanoparticles from the roadside environment is effectively prevented by the existing topography of the region (Cusack et al., 2013). The topography includes

natural and artificial means such as vegetative and building environment. The vegetative barrier is the cost-effective method and showed a significant reduction of particle number concentration in the nearby micro environment (Nøjgaard et al., 2012). Various research studies found that the vegetative barrier in the roadside microenvironment acts a barrier for the air pollutants preventing it from dispersing in the nearby environment (Mohan et al., 2024). The role of vegetative barrier on nanoparticles reduction is still a grey area to research and having lot of potential scope in future. Very few studies are conducted in this direction and found interesting results (Dasappa and Camacho, 2021). During a low wind condition of <0.5 m/s the concentration of the particle number concentration reduced to around 37.7 to 63.6 % due to the vegetative barriers in the downwind condition (Ghei and Sane, 2018). Similarly, another study found that the efficiency of the vegetative barrier depends on the thickness of the vegetative barrier (Kanawade et al., 2014). Based on the wind direction and speed the efficiency of the vegetative barrier varies and it not only helps in reducing the concentration of the nanoparticles but also plays a major role in reducing the Respiratory Deposition Dose of an individual from 36 to 80% (Zheng et al., 2021). The results of all the studies suggest that the efficiency of the vegetative barrier increase when they are placed at the downwind condition (Fangqun, 2010). The performance efficiency of the vegetative barriers is found high when compared with the solid barriers such as buildings due to the higher surface area of the vegetation which helps in dry deposition of the nanoparticles (Fatima et al., 2022). Studies revealed that the vegetative barriers installed on the roadside environments are not only successful in reducing the particle number concentration of the nanoparticles but also useful in reducing the co pollutants such as BC which also emitted in higher concentration from the transportation sector in the urban regions (Mohtar et al., 2018).

# 2.11 Nanoparticle measurement and detection techniques

Nanoparticles emitted in the atmosphere have wide range of shape, structure, size distribution and composition (Manigrasso et al., 2020). Measurement of this nanoparticles are challenging but due to the recent advancements in instrumentation there are several instruments with different principles are found for the measurement of these nanoparticles (Hu et al., 2017). The concentration of the nanoparticle's

changes with distance from the source in this regard the concentration of the nanoparticles inside the engine exhaust is more and the temperature of the plume will be high even the measurement of PNC in that condition is also possible due to the advance instruments (Almeida et al., 2015). The European commission provides the regulation for the vehicle exhaust emission standards (Gómez-Moreno et al., 2011). In nanoparticle measurement techniques measurement through aerodynamic diameter and electrical mobility are the two most commonly used measurement technique by the researchers globally (Jianhua et al., 2005). However, the efficiency of the measurements changes from manufactures to manufactures and from one principle to different principle (M. Tiwari et al., 2014). There major parameters of the nanoparticles which needs to be measured are concentration, size distribution and physiochemical properties (Casquero-Vera et al., 2022).

# 2.11.1 Gravimetric Concentration measurement

The concentration measurement of the nanoparticles in the urban regions are done for both mass and number concentration (Kumar et al., 2008b). The mass measurements of the nanoparticles are done using a different types of cascade impactors where the filter papers are used for the collection of the dust particles (Herr et al., 2021). The mass measurement provides a specific size cut off range for a different range of size bins which are arranged stage wise allowing air particles to enter each stage and get collected on the impactor (Patel et al., 2021). The instruments use a different size cut off range based on the different manufacturer and the even customisable size ranges and a single pump for maintaining the air flow in the impactor (Dumka et al., 2018). The cascade impactor provides non-real time data and statistically not significant for lesser sampling period (Sarangi et al., 2018). Conventional impact factors are not favourable for the measurement of particles  $< 0.4 \mu m$ . Dekati Low pressure impactors are used for the measurement of particle size ranges from 10 nm to 10µm. Micro orifice uniform distributed impactor with a wide sample flow rate of 10 to 100L/min are capable of measuring particles from >30 nm to <2500 nm. The impactor collects the dust on a quartz plate which can be further used for physiochemical analysis (Li et al., 2013).

#### 2.11.2 Optical measurement

In the detection technique of atmospheric nanoparticles employ light beam scattering, absorption principles (W. C. Lee et al., 2014). The light scattering and absorption measurements are used for real time measurements. The light scattering method in which the scattered light is absorbed at the receiver (Buonanno et al., 2014). The deviation of the light sources in the receiver end is based on the particle size present in the pathway (Baldauf et al., 2013). When the light source is emitted in the measurement chamber the particles present in the chamber scatters the light the angle of scatter and the photometer detects the scattered light (Ma and Birmili, 2015). The number concentration of the particles is detected through the count rate and the size is determined through the pulse height variation (Weichenthal et al., 2016). Commercially available light scattering based instruments can detect particles from few nanometres to several hundred micro meter particles with different measuring angles such as 30°,  $45^{\circ}$ , > 90° based on the technology used (Tiwari et al., 2018). The instruments require proper maintenance and cleaning of the particle counting chamber for more reliable data (Babu et al., 2016). The optical particle counters and condensation particle counters are widely used light scattering technique instruments in atmospheric nano particle measurement instrument. The condensation particle counters (CPC), based on this light scattering principle is the most commonly used instrument for the lower size nano particle measurements such as ultrafine particles. However, the atmospheric nanoparticles have a detection limit due to its lower size so there is a condensation chamber is used to grow the size of the nanoparticles using a condenser (Cong et al., 2017). The condensation solution such as butanol or distilled water is used for condensation. The other size ranges such as quasi-ultrafine particles are measured through the optical particle counters (Ezz et al., 2015). The CPC combined with a differential mobility analyser (DMA) is used for the effective size distribution measurement of the atmospheric nanoparticles (Reggente et al., 2015). The light absorption techniques are used for black carbon (BC) based ultrafine particles due to their strong correlation. The instruments such as Photoacoustic Soot Sensor (PASS) and laser induced incandescence (LII) are advanced instruments for measuring the BC based nanoparticles from exhaust or tailpipes.

# 2.11.3 Microbalance techniques

The ultrafine particles collected through the mass measurements are collected on surface of an oscillating element that occurs due to the change in resonance frequency (Meier et al., 2015). The microbalance is designed for the measurement of the sampled nanoparticles through quartz (Simon et al., 2017). The sampled particles are deposited on the thin quartz crystal resonator. The decrease in resonance frequency is used for the measurement of particle mass. The ambient temperature and humidity influence the concentration of the measurement (Padró-Martínez et al., 2012). The instrument mostly uses different size stages for the collection of these particles. The measured nanoparticles are further used for physiochemical analysis (Lee et al., 2012).

# 2.11.4 Microscopic technique

Microscopy techniques are used for examine the morphology of the measured particles (Gordon et al., 2012). The particles are collected on specific filter paper in the cascade impactor or MOUDI instruments are then magnified for microscopic analysis (Carpentieri and Kumar, 2011). The instruments such as TEM, SEM and TCP-MS are used for the wide range detection of particle morphology, radius, dimensions, and other morphological property of the particles (Fuller et al., 2012). In recent technological development there are few real-time measurements of nanoparticles coupled with TEM are also available(Cheng et al., 2019). The drawback of this technique is that this technique is suitable for solid particle analysis but not suitable for semi volatile particles due to the particle evaporation or loss during vacuum and electron beam heating.

# 2.11.5 Impactors technique

The cascade impactors woks of the principle of particle inertial classification. The nanoparticles are passed through series of stage filters through an array of nozzles above a solid substrate. The nanoparticles which are having aerodynamic diameters greater than the design cut point are impacted where the smaller particles follow the air flow and got deposited on different stages of impactor based on the cut diameter. The impactors efficiency is proportional to strokes number. Due to the inertia the smallest particles which pass through the final stage of the cascade impactors are not

collected. The advanced cascade impactor can collect nanoparticles from 30 nm to 10  $\mu$ m. some of them are low pressure cascade impactors, Dekati low pressure cascade impactors are fee examples. The Electrical low-pressure Impactor (ELPI), is a real time high-tech impactor in which unipolar corona charges are used along with inertial classification and electrical detection. The Dekati electrical low-pressure impactors have minimum scale time of 1s with a capacity to measure particles from 7 nm to 10  $\mu$ m.

# 2.11.6 Mobility analyser

The differential mobility analyser and scanning mobility particle sizer are the two majorly used mobility. In differential mobility analyser the sampled air is passed through a central rods outer boundary which is made up of a series of electrically isolated rings that's are linked to electrometers (Noble and Hudson, 2019). The size distribution of nanoparticles is measured by the deposition of charged particles in the ring electrode. The DMS has lower particle sensitivity than the scanning mobility particle sizer (SMPS). The sampling time resolution of DMS is (0.1 to 1Hz) so it is more commonly used in vehicle exhausts emission monitoring and dilution.

The Fast-integrated mobility spectrometer consists of a charger, size classifier, condenser and detector. The sampled particles pass through a neutralizer thereby receiving a charge distribution of bipolar equilibrium. Based on the electrical mobility charge the particles are separated into different paths. These particles are then passed through a supersaturated butanol vapor where the particle size increases their size by condensing over them (Jbaily et al., 2022).

# 2.12 Modelling of nanoparticles

Land use regression (LUR) models are widely used model for distribution of spatial variation of nanoparticles and aerosols (Tawiah et al., 2022). This model uses the air pollution concentration of different pollutants at different locations which spread over a larger area and they develop stochastic models through prediction functions available in GIS (Farmer et al., 2019). AERMOD is another widely used model for simulating simultaneously different shapes, characteristics and elevations based on the gaussian dispersion theory. The ADMS5 models are widely used models by researchers, which

is capable of calculating concentration of pollutants generated from all the three different sources (point, line and area sources) and maximum of 300 sources can be handles through this model at a time (Benka-Coker et al., 2020). ADMS and AERMOD have similar algorithm and the dedicated roadside pollutants are modelled using ADMS-Roads software (Singh et al., 2021). OSPM is the dedicated model used for ultrafine particles in Denmark developed by the National Environmental Research Institute. The OSPM model is used for modelling the ultrafine particles in the urban street canyon based on the turbulence and meandering effects of winds (Pokhrel et al., 2015). WRF-Chem is another widely used atmospheric chemistry models in the aerosol science for simulation of aerosols, trace gases and particulate matter. The model has two modules such as weather module and chemistry module which can be used for stimulating meteorological and chemical parameters on the different pollutants simultaneously (Kumar et al., 2022). The WRF-Chem models provides the more realistic and accurate results for aerosol prediction. During the year 2000, MONO-32 and MULTIMONO model are developed for the nanoparticle modelling. It was developed to study their properties and the use of this model is that, it can investigate the role of different emissions in determining the composition of the different sources. The model considers both gaseous and particulate emissions including the gas phase chemistry for modelling (Kesarkar et al., 2007). The model include parameters such as dry deposition loss, nucleation, condensation and interintra molecular coagulation parameters also. The models can be used in combination with the computational fluid dynamics (CFD) models to form an advance sophisticated model such as MPPD (Manojkumar et al., 2022). USEPA developed different models to determine the concentration of the atmospheric nanoparticles specially the R-line models are used for the roadside emissions (Zou et al., 2024). The R-line model helps in estimating the near road emission dispersion and its exposure to humans. This model is most suitable for the exposure risk assessment and also designed to predict pollutant concentration at the receptor side which is located adjacent to the road. The main advantage of this model is that this model have an embedded wind meander algorithm that helps to predict the dispersion of the pollutants in all the different directions based on the light and variable winds (Estévez-García et al., 2020). The GAM models are widely used model the prediction of atmospheric nanoparticles. The GAM model was developed by Hastie and Tibshirani which was latter modified by the Wood (Steenland et al., 2018). This model is a hybrid model which have features of both LUR and R-line model. In this model the time series estimation and spatial characteristics while representing the nonlinear relations are widely used by different researchers (Lorelei de Jesus et al., 2020). Previously the GAM models are used for the single point sources and mobile monitoring sources with less temporal coverage but in recent days with the advancements in the model helps to estimate the different process of the atmospheric nanoparticle process such as nucleation, deposition, dilution and long-range transportation. The Artificial Neural Networks and machine learning tools are used for modelling large data sets with different parameters along with the atmospheric nanoparticles (Jung et al., 2023; Xu et al., 2020; Wu et al., 2024).

#### 2.13 Health impacts of nanoparticles

The health impacts due to the exposure of atmospheric nanoparticles are developing research area (Salvi and Apte, 2016). Very few epidemiological studies are done in examining the mortality and morbidity of the atmospheric nanoparticles (Kwon et al., 2020). The atmospheric nanoparticles are smaller in size, mass and their deposition and clearance are totally different from larger size particles (Figure 2.3) (Singh et al., 2019). The deposited particles in the human body is governed by diffusional mechanism and the atmospheric nanoparticles are expected to translocate within the body due to their smaller size (Deng et al., 2019). The deposition estimation of the atmospheric nanoparticles is expected to be in the alveolar region and they are potential to enter the blood stream (Shupler et al., 2024). Very few experimental studies proved the translocation of atmospheric nanoparticles in the blood streams of the human body. Based on the exposure time the health impacts of the atmospheric nanoparticles are determined (Saha et al., 2024). Short term exposures to atmospheric nanoparticles can cause cardiovascular health problems as well. The atmospheric nanoparticles after inhalation can penetrate deep into the respiratory system such as lung tissues. Recent studies revealed that the human alveolar macrophages can't able to remove the particles less than 70 nm causing the particles to translocate in the blood streams (Rentschler and Leonova, 2023). The translocated nanoparticles are transferred from the lungs to the liver and their depositions are found in the liver after the exposure. The exposure to these atmospheric nanoparticles also causes sequential lung inflammation (Velasco and Tan, 2016). Few epidemiological studies (Kwon et al., 2020; Xu et al., 2021; Yu et al., 2022) showed inconsistent results on comparing the particle exposure and mortality rate.

Recent study in UK estimated an average loss of about 6 months life expectancy occurs due to the resident's exposure to  $PM_{2.5}$  and such studies are not found extensively for the nanoparticle's exposure (Morales Betancourt et al., 2017). The toxicity of the atmospheric nanoparticles due to their characteristics such as shape, concentration, size distribution, surface area, volume and other characteristics such as chemical composition, surface chemistry, surface charge and crystal structure are hard to find the suitable metric to quantify the human exposure (Joodatnia et al., 2013). Few studies suggest the surface area is the suitable metric for health impact assessment and few suggest the concentration (Bouma et al., 2023). The particles with higher surface area to mass ratio have great contact area for materials to interact with biological surfaces (Manojkumar et al., 2021). Majority of the toxicology studies proved that the number and surface area pose a great effect on human health.

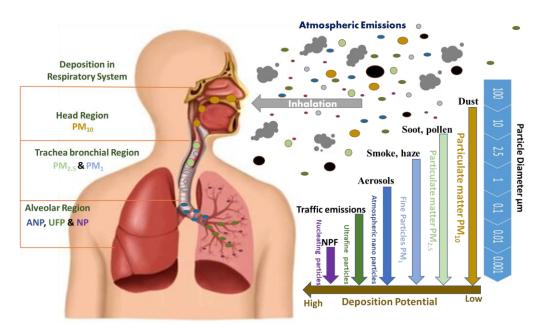


Fig 2.3 Deposition of different size pollutants in the respiratory system

Due to lack of long-term studies such as  $PM_{2.5}$  there is very less evidence of nanoparticles health effects. The atmospheric nanoparticles have great potential for

harming health but its precise function is still unknown for many diseases (Goel and Kumar, 2015). The roadside environment in the urban region is one of the major hotspots for atmospheric nanoparticles which is having higher concentrations compared to the background regions. Many developing cities are found in the Asian regions compared to Europe and united states, that have higher potential risk due to their high number concentration. The children's and old age people are more vulnerable to the exposure of these nanoparticles which causes respiratory and cardiovascular illness (Gould et al., 2020).

The deposition efficiency of the nanoparticles in the human respiratory system are not uniform due to their different particle sizes. The higher size particles are deposited in the head air way region and the lower size particles are deposited in the alveolar region (Manojkumar and Srimuruganandam, 2021b). The atmospheric nanoparticles from the traffic emissions are deposited in the epithelium cells and then transported to the central nervous system such as olfactory bulb which can cause Alzheimer's disease and other brain disease (Gabdrashova et al., 2021). The transport emissions are responsible for the decreased cognitive function in elderly people (Zhu et al., 2016). The reduced particle size and increased surface area has adverse toxicity in air pollution related health effects (Delapena et al., 2018). Due to their higher surface area the atmospheric nanoparticles can have high capacity to absorb the organic gases and heavy metal elements as well which grows further due to the agglomeration. The particles with low mass, high surface area ratio show different physiochemical properties (Bergmann et al., 2021). The unique properties of the atmospheric nanoparticles make its toxicity more complicated compared to the toxicity of other materials (Manojkumar and Srimuruganandam, 2021a). The nanoparticle exposure in the human body is mainly governed through the respiratory exposure. Respiration and adsorption are possible ways of exposure. The other ways of exposure to atmospheric nanoparticles are Respiratory tract, Derman exposure and ocular Exposure.

The atmospheric nanoparticles enter through the nasal cavity through inhalation. The inhaled nanoparticles enter the downstream airway of the lungs therefore the UFP get more interaction with lungs including trachea, bronchia and alveolus (Arub et al., 2021). The interaction of these particles goes beyond the tissue cell with mucosa. The

lungs do not have the ability to clean or filter the fine or ultrafine particles through its filtering mechanism. The human nose has a limitation for filtering atmospheric nanoparticles (Schwarz et al., 2023). The filtering capacity of the human lungs is 80% for the 1µm particles and less than 5% for particles having diameter around 100 nm during the resting period (Mosonik et al., 2019). The lungs are the most important part of the respiratory system which have direct contact with the atmosphere. The two major parts of the lungs are the airways and the alveolar structure. The alveolar region contains a monolayer epithelial cell which can increase the possibilities of the atmospheric nanoparticles entering into the blood gas barrier (Badami et al., 2024). The nanoparticles are not capable of penetrating the light junction of the cells due to their size but can enter through the epithelial cell body. It can also be transported into the cell cytoplasm due to the passive diffusion. For higher size particles the particles enter into the cell directly. The nanoparticles after crossing the blood-gas barrier enters the blood circulation and gets contacted with the extra pulmonary tissue cell. The cell interaction or deposition of the nanoparticles can cause reactive oxygen species (ROS). The circulation of the nanoparticles in the human body increases it toxicity and inflammation compared to the other size particles (Klemm et al., 2022).

The alveolar region has the highest deposition efficiency for the smaller size particles. The particles with diameter 20nm are found to have deposition potential of about 50% and the particles having concentration > 100 nm to 2.5µm are found to have 10 to 20% efficiency (Vu et al., 2017).

Multiple path particle dosimetry model is one of the sophisticated models for the estimation of nanoparticles deposition in the different regions of the lungs (Ma et al., 2022). The model is useful for estimating the dose of deposition. Ocular exposure is another possible way of nanoparticles exposure in the human body. The exposure occurs due to the floating of particles or through rubbing of eyes in the area. The nanoparticles can reach to the brain cells. Apart from alveolar deposition the olfactory bulb deposition is another issue for the human health risk assessment through the atmospheric nanoparticles.

# 2.14 Studies related to the proposed study area

Location	Туре	Size	Num.		Summary
		range	Con.	Reference	
		(nm)	(x10 <sup>3</sup>		
			#/cm <sup>-3</sup> )		
New	Urba	10-	42 to 24	Rajagopal	This study measured the
Delhi,	n	1000		et al., 2023	atmospheric nanoparticles
India	roads	nm			concentration in the
	ide				roadside environment
					during different emissions
					scenarios which shows the
					direct impact of
					transportation sector in the
					atmospheric nanoparticles
					in the road side
					microenvironments.
New	Urba	10-	41 to 24	Rajagopal	The study measured the
Delhi	n	1000		et al., 2024	concentration of the
	roads	nm			atmospheric nanoparticles
	ide				in the roadside
					environment for different
					seasons showing winter
					season as the highly
					polluted season comparted
					to others.
New	Urba	10-	26-day	Mohan et	The study conducted for
Delhi	n	1000	time	al., 2024	the different hours of the
	roads	nm	24 night		day such as day time and
	ide		time		nigh time based on the

Table 2.3 Summary of the works done so far in the proposed study area

					vehicular intensity and
					source variation.
New	Urba	10-	2.9 in	Mohan et	The study conducted in
Delhi and	n	1000	backgrou	al., 2024	different geographical
Rani-	regio	nm	nd and		variation such as urban
chouri	n and		25 in		region and background
	back		urban		conditions.
	groun		regions		
	d				
	regio				
	n.				
New	Insid	5 nm	14.8	Jose et al.,	This study site is
Delhi,	e	to		2021	dominated by the Aitken
India	urban	32µm			mode particle, which
	area				follows a bimodal diurnal
					variation pattern, with
					peaks during morning and
					evening traffic hours.
					Accumulation mode
					particles are scarce over the
					study location except
					during the winter and post-
					monsoon days, when it
					shows a bimodal secondary
					peak along with the
					primary Aitken mode
					particle.
New	Insid	10 to	25.52	Kanawade	New particle formation
Delhi,	e	800		et al., 2020	events study was
India	urban				conducted and reported
	area				higher GR rate with a

					higher concentration in
					Aitken and accumulation
					mode particles.
IITK,	Insid	4.45 to	10.195	Kanawade	The study emphasis the
Kanpur,	e	736.5		et al., 2020	secondary aerosol
India	urban				formation process in the
	area				highly polluted urban
					regions.
New	Insid	4 to	-	Sarangi et	New particle formation or
Delhi,	e	661		al., 2018	gas to mparticle conversion
National	urban				process observed during
Physical	area				the night time in Delhi city.
Laborator					Which is a first of its kind
у					in the study region.
Ooty,	Торо	10.9 to	2.845	Kompalli et	The study analysed the
India	graph	461.4	±1.184	al., 2018	concentration of the
	ically				aerosol particles in the
	300				background region in the
	m				country and its size
	above				distribution.
	the				
	valle				
	У				
Mahabale	Back	5 to	21.800	Leena et al.,	In tropical regions, such
shwar,	groun	1000		2017	studies are limited,
India	d				especially in the Indian
Physics	1348				region
Laborator	m				
У	above				
	mean				

	sea				
	level				
Trivandru	Back	15 to	~5.119	Babu et al.,	The study identified the
m,	groun	15,000	(±3.139)	2016	UFP particles burst during
India	d				two different periods. First
	(10K				period is during the onset of
	М				land breeze event in
	from				evening or midnight. The
	urban				seond event during the
	area)				sunrise hours in the seas
					breeze.
New	Insid	9 to	37.2(eve	Sarangi et	New particle formation
Delhi,	e	425	nt day	al., 2015	study
India	urban		24.9(Non		
National	area		-event		
Physical			Day)		
Laborator					
У					
Trivandru	Back	10 to	~15.900	Kompalli et	New particle formation
m,	groun	875		al., 2014	study during late evening
India	d				and nigh hours
	(10K				
	М				
	from				
	urban				
	area)				
Hanle,	Back	16 to	~2.700	Kompalli et	New particle formation
India	groun	1364		al, 2014	event studies
	d				

	(~30				
	0m				
	above				
	the				
	surro				
	undin				
	g				
IITK,	Insid	4 to	-	Kanawade	The particle number
Kanpur,	e	750		et al., 2014	concentrations were found
India	urban				to be less in Kanpur
	area				resulted in larger particle
IITM,	Outsk	14 to	-	Kanawade	sizes creating higher
Pune	irt of	750		et al., 2014	condensation sinks. The
India	Urba				mean particle mode
	n area				diameter at Kanpur was
					larger by a factor of ~1.8
					than at Pune. Generally, the
					particle growth rates were
					higher at Kanpur, whereas
					the formation rates were
					higher at Pune.
IITM,	Outsk	0.46 to	-	Siingh et	Positive and Negative ions
Pune	irt of	50		al., 2013	measurement
	Urba				
	n area				

# 2.15 Summary

The detailed literature review provides information about the global studies and also the research gaps which needs to be addressed in further studies.

# CHAPTER-3 METHODOLOGY

# 3.1 Introduction

The air quality in the global south region are deteriorating at an alarming rate. The majority of the growing urban region are located in the global south region especially in the Asian subcontinent. The rapid urbanization in these regions are responsible for the poor air quality which are generated through various process such as emission from industries, transportation sources and other sources. The air pollution in the urban regions is a complex mixture of different sources, among them transportation sources is one of the major emission sectors. The current study was conducted in the Indian subcontinent which is one of the highly populated country. Delhi city is capital of India with a population of about 33.81 million people. The population in Delhi increases at a rate of 2.63% compared to the previous year (2023). The city experiences severe pollution events and ranked among the highly polluted city globally is the proposed study region (Sharma et al., 2025). Due to its geographical location and climatic conditions the capital city experiences different pollution events. The air pollution related human health risks in the city is estimated to be around 7 million premature deaths. The Delhi city having an area of around 1483 km<sup>2</sup>. The vehicle population in about 11 million and it is predicted to increase to 25.6 million by 2030. The personnel vehicle population in Delhi accounts for 2.9 million for cars/jeeps and 6.1 million twowheelers. Two wheelers and cars alone contribute for around 93% of the total vehicle population in Delhi (Rajagopal et al., 2023). So, the emission from the transportation sector is one of the major sectors for the atmospheric nanoparticles emission in the Delhi city which is considered as the study region for the study.

## **3.2 Study location characteristics**

The monitoring site is located in the North West region of Delhi city (28.75° N, 77.12° E). The region geographically falls under the Rohini District (Fig. 3.1). The measurements are done using a real-time air quality monitoring station which is located adjacent to the Delhi Technological University campus. The study aims to measure the emissions from the transport sector in the road environment so the monitoring station is placed adjacent to the road. The road where the measurements

are done is classified as state highway without central meridian. The road is a major connecting road for Delhi and Rohtak city which is located in the nearby state Haryana. The vehicle density in the selected road is around 1300 vehicles/ hr with a daily average of about 3500 vehicles. The vehicle density in the road includes different types of vehicles such as cars personnel and commercial, three wheelers, E-rickshaws, light motor vehicles, heavy motor vehicles, two wheelers, E-busses. The major commercial fuels used by the vehicles are petrol, diesel compressed natural gas and electricity.

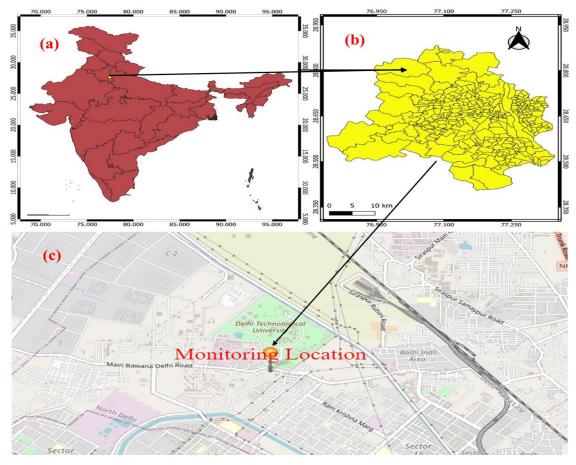


Figure 3. 1: (a) map of India with all the states of the country, (b) map of Delhi city with major road network, (c) geographical location of the study area

The sampling of the nanoparticles is done at distance < 2m from the road. The sampling was done adjacent to the pedestrian way in the road. The major sources of pollution in the study region includes waste burning, transport emissions, road dust and other sources. The location of the monitoring station provides way for the monitoring station to receive the emission from the transportation sector majorly. The study region is surrounded by different types of primary and tertiary roads.

The study region is surrounded by educational institutions, residential areas and road networks.

The region is well connected through road, rail and metro connectivity. The metro station and the railway station are located in the south east direction of the monitoring station with an aerial distance > 2km (Figs. 3.2, 3.3). Within the aerial distance of 1 km a major bus depot is located and the Indira Gandhi international air post located at the southern direction of the monitoring station. The north direction is surrounded by residential area and road network. The study region comes under the Indo Gangetic Plain region. This Indo Gangetic Plain region is known for the highly polluted region with majority of the highly urbanising cities lies in this region. The mighty Himalayan Ranges acts as barrier for the air passage in this region which resulted in accumulation of pollutants in this region.

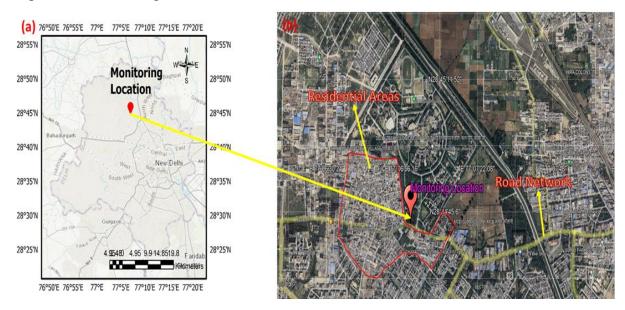


Figure 3.2: (a) Map of Delhi showing the study area, (b) Aerial view of the monitoring site with the surroundings



Figure 3.3. Picture of the monitoring station and its location

# 3.3 Measurement period

The measurements of the different pollutants are done for a year starting from December 2020 to November 2021 (Fig 3.4). This year long measurement includes all the major seasons of the study area and also the major events in the year. The study period also includes a emission restriction period as well which helps to identify the role of transport emissions in the study area. The data collection was done for 298 days in a year out of 365 days.

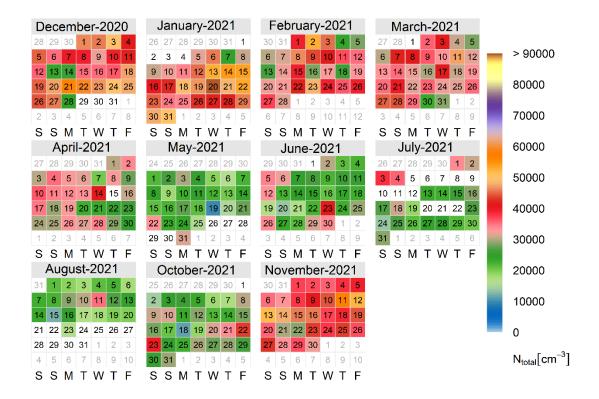


Figure 3.4 calendar plot showing the monitoring days with total PNC concentration on particular day

Based on the objectives, the study period is classified into two different types seasonally and also based on the emission episodes. The episodic classifications are based on the restriction imposed for the movement of public during the outbreak of corona virus in the year 2021. The emission episodes are classified into two major period namely, period I and II. Both the periods have three different phases: phase I will have normal scenario, phase II have an episodic event either less emission or extreme emission and phase III will be the period after the episode. These are the three sub classifications in both the periods. The period I consist of total 91 days in which the first phase consists of 20 days, 47 days in second phase and 24 days of monitoring in the third phase. Similarly, in the period II, phase I consist of 33 days, phase II consists of 8 days and phase III consist of 18 days with an overall data of about 59 days. The detailed classification of episodes and their respective phases are depicted in the Table 3.1.

Sr.	Period	Time period	Phase	<b>Total Days</b>
no				
1	Ι	1-20 April 2021	BR- Before Restriction	20
2	Ι	21 April – 5 June 2021	DR- During Restriction	47
3	Ι	6-30 June 2021	AR- After Restriction	24
4	II	3 October-4 November 2021	BD- Before Diwali	33
5	II	5-12 November 2021	AD- After Diwali	8
6	II	13-30 November 2021	DR-II- During Restriction II	18
		Total Days		150

Table 3.1 Classification of different periods based on the imposed restrictions.

The percentage of vehicular reduction obtained during the period I and II is taken from the google mobility data of this particular region. The google mobility data provides the information about the vehicular density changes during the different phases of the year. The north west Delhi region is selected for the google mobility data analysis.

The measurement period is further classified seasonally based on Indian science academy classification. The study region has five major seasons namely winter, spring, summer, monsoon and autumn. The climatology of the study region varies greatly during different season from extreme hot summers to cold winters. The climatic conditions play a major role in determining the concentration and other secondary transformations of these atmospheric nanoparticles. The detailed time frame and classifications of the season along with the information of data available for the particular period is mentioned in the Table 3.2.

Table 3.2 Seasonal classification of the study period.

Seasons	Time period	Total number of days	Monitored days
Winter	December to mid- February	76	69

Spring	Mid-February to	45	44
	March		
Summer	April to June	91	84
Monsoon	July to mid-	77	41
	September		
Autumn	Mid-September to	76	60
	November		
Tota	l days	298	365

# **3.4 Instrumentation**

Various instruments are used in the study for measurement of different types of the pollutants. All the instruments are kept in a fully airconditioned monitoring station. The monitoring station consist of a separate sampling channel for nanoparticles measurement and gaseous measurement. The monitoring station is equipped with continuous powers supply and the Realtime measurements are done and the collected data are stored in a computer for all the measurement. The pictures of the monitoring station are shown in the figures (Figs. 3.5, 3.6).





Figure 3.5 Picture of the monitoring station, located adjacent to road

Figure 3.6 Internal view of the monitoring station

#### 3.4.1 Atmospheric nanoparticle measurement.

The atmospheric nanoparticles are measured using a scanning mobility particle sizer (SPMS +C) with condenser. The SMPS+C instrument was manufactured by GRIMM aerosol Technik GmbH & Co company. The serial number of the instrument is 54031801 (Figs 3.7,3.8). The instrument consists of a neutralizer, differential mobility analyzer and a condensed particle counter combined together. The instrument uses a Vienna type L-DMA which has capacity to measure particles from 10nm to 1090nm. The air is sampled through a diffusion dryer which helps to maintain the relative humidity of the air sample and then the sampled air is passed through a neutralizer made of made of radioactive material (Am-241). The air is then passed through a DMA consist of a length of 0.35 m. The inner diameter of the DMA is 0.040 m and outer diameter is 0.026 m. The optimum DMA temperature is 20° C and pressure is 101300 Pa. The slip factor of the instrument used is 2.83 and the effective length of the sampling probe without losses is 0.38m. The time scale for one complete cycle of measurement is ~7 minutes and the different size particles are distributed in a 45 different size segregated channels.

The size distribution calculation of the different size bin calculations is based on the Winklmayr et al. (1991) 9(d) equation and it follows the ISO 15900 standards. The size distribution calculations flows the different mathematical equations that are described below.

$$N_c(D) = \left(\frac{Q_a}{Q_{sh}}\right) \cdot \sum_{i=1}^{\infty} [\alpha(D,i) \cdot \Psi(D) \cdot F(D)]_{D=D_i} - (1)$$

Where,  $N_c(D)$  – Total number concentration measured in CPC

D- Diameter of single charged particle that corresponds to voltage of DMA.

D<sub>i</sub>- Diameter of the particle with i charges that correspond to voltage of DMA.

Q<sub>a</sub> – Volume flow rate of the sample airflow.

Q<sub>sh</sub>-Volume flow rate of the sheath airflow.

i- Number of elementary charges carried by the particle.

α (D<sub>i</sub>, i)- probability of the particle to carry i elementary charges (Fuchs-Wiedensohler values).

 $\psi$  D<sub>i</sub> – function converts diameter interval into mobility interval.

F(D)- Number size distribution at the inlet of DMA (Rajagopal et al., 2023). Two factors are added to this formula, one is eff CPC (D) which is accountable for the efficiency of the CPC particles having size lesser than 10nm; and the second one is eff CPC (D) which considers the particle loss in the DMA. The particles with 10 different charges can be neglected for the practical assumption. Then the calculation is as follows

$$N_{c}(D) = \left(\frac{Q_{a}}{Q_{sh}}\right) \cdot \sum_{i=1}^{10} [eff \ CPC \ (D). \ eff \ DMA(D). \alpha(D, i). \psi(D). F(D)]_{D=D_{i}} - (2)$$
  
By solving this equation  $F(D)$  is

By solving this equation  $\Gamma(D)$  is

$$F(D) = \frac{1}{eff_{CPC}(D).\ eff_{DMA}(D).\ \alpha(D,i).\ \psi(D)} - (3)$$
$$\{\frac{\frac{N_{c}(D).Q_{sh}}{Q_{a}} - \sum_{i=2}^{10} [eff_{CPC}(D).\ eff_{DMA}(D).\ \alpha(D,i).\ \psi(D).\ F(D)]_{D=D_{i}}\} - (4)$$

The standard stepping mode  $F(D)_{D=Di}$  can be exactly calculated for i =2 from the previously measured size channels.



Figure 3.7 Picture of the scanning mobility particle sizer

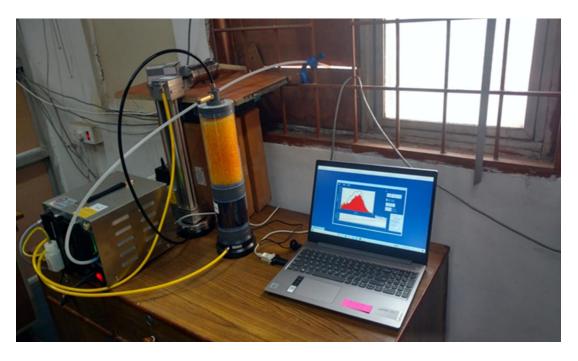


Figure 3.8 Picture of the scanning mobility particle sizer

#### 3.4.1.1 Number distribution calculation

The different size distribution calculations are based on the following mathematical calculations given below.

The values of the size distribution of the particles  $(n_i)$  for the size interval (i) was expressed in the ratio of the absolute concentration. N<sub>i</sub> of this interval and the different size range  $\Delta D_p$ . Then the equation is

$$N_i = n_i \Delta D_p - (5)$$

The smaller size bin taking the limit  $\Delta D_p \rightarrow 0$  the limit  $\Delta D_p$  becomes small and the values equals to  $dD_p$ . Then the size distribution are as follows  $n_N(D_p)$ .

$$n_N(D_p)dD_p = \text{number of particles per cm}^3$$
  
 $D_p \text{ to } (D_p + dD_p) - (6)$ 

The units of  $n_N(D_p)$  are  $\mu m^{-1}$  and the total particles measured in per cm<sup>3</sup>. Then N<sub>t</sub> is then

$$N_t = \int_0^\infty n_N (D_p) dD_{p-(7)}$$

The cumulative size distribution function  $N(D_p)$  is defined as

$$N(D_p)$$
 = number of particles per cm<sup>3</sup>.

The actual particle in the size range

$$N(D_{p}) = \int_{0}^{D_{n}} n_{N}(D_{p}^{*}) dD_{p}^{*} - (8)$$

Finally, the size distribution can be expressed as

$$n_N(D_p) = dN/dD_p - (9)$$

#### 3.4.1.2 The surface area, volume and mass distributions calculation

The properties of the measured particles properties changes based on the particle surface area, volume and mass distributions of the particles.

The total surface area of the particle measured are obtained from the equation

$$S_{t} = \pi \int_{0}^{\infty} D_{p}^{2} n_{N} (D_{p}) dD_{p} = \int_{0}^{\infty} n_{S} (D_{p}) dD_{p} \quad (\mu m^{2} \ cm^{-3}) \ - (10)$$

Similarly, the volume concentrations are based on the

$$V_t = \frac{\pi}{6} \int_0^\infty D_p^3 n_N (D_p) dD_p = \int_0^\infty n_V (D_p) dD_p \quad (\mu m^3 \ cm^{-3}) \ -(11)$$

The mass of the measured particles is obtained using the following calculation.

$$n_M(D_p) = \left(\frac{\rho_p}{10^6}\right) n_V(D_p) = \left(\frac{\rho_p}{10^6}\right) \left(\frac{\pi}{6}\right) D_p^3 n_N(D_p) \quad (\mu g \ \mu m^{-1} \ cm^{-3}) - (12)$$

The mean values used in the size distribution are as follows (Table 3.3)

Table 3.3 Different calculations of the nanoparticles based on number measurements

Property	Defining relation
Number mean diameter $\bar{D}_p$	$\bar{D}_p = \frac{1}{N_N} \int_0^\infty D_p n_N (D_p) dD_p$
Median diameter D <sub>med</sub>	$\int_{0}^{D_{med}} n_N \left( D_p \right) dD_p = \frac{1}{2} N_t$
Mean surface area $\overline{S}$	$\bar{S} = \frac{1}{N_p} \int_0^\infty n_S(D_p) dD_p$
Mean volume $\bar{V}$	$\bar{V} = i_{N_1} \int_0^\infty n_v (D_p) dD_p$
Surface area mean diameter <i>D<sub>S</sub></i>	$N_t \pi D_S^2 = \int_0^\infty n_S(D_p) dD_p$
Volume mean diameter	$N_t(\pi/6)D_V^3 = \int_0^\infty n_V(D_p)dD_p$

Surface area median diameter $D_{S_m}$	$\int_{0}^{D_{Sm}} n_{S}(D_{p}) dD_{p} = \frac{1}{2} \int_{0}^{\infty} n_{S}(D_{p}) dD_{p}$
Volume median diameter $Dv_{\rm m}$	$\int_0^{D_{vm}} n_v(D_p) dD_p = \frac{1}{2} \int_0^\infty n_v(D_p) dD_p$
Mode diameter <i>D</i> <sub>mode</sub>	$\left(\frac{dn_N(D_p)}{dD_p}\right)_{D_{\text{mat}}} = 0$

#### 3.4.1.3 Properties of lognormal distribution

The lognormal distribution of the particles is based on the following calculations.

$$n_{S}(D_{p}) = \frac{\pi D_{p}^{2} N_{t}}{(2\pi)^{1/2} D_{p} \ln \sigma_{g}} \exp\left(-\frac{\left(\ln D_{p} - \ln \bar{D}_{RS}\right)^{2}}{2\ln^{2} \sigma_{g}}\right) - (13)$$

$$n_{S}(D_{p}) = \frac{\pi N_{1}}{(2\pi)^{1/2} D_{p} \ln \sigma_{g}} \exp \left(2 \ln \bar{D}_{PS} + 2 \ln^{2} \sigma_{s}\right)$$
$$\times \exp \left(-\frac{\ln D_{p} - \left(\ln \bar{D}_{PS} + 2 \ln^{2} \sigma_{g}\right)\right]^{2}}{2 \ln^{2} \sigma_{g}}\right) \quad - (14)$$

#### 3.4.2 Gaseous pollutant measurement

The gaseous pollutants such as carbon monoxide (CO), ozone (O<sub>3</sub>), Sulphur dioxide (SO<sub>2</sub>) and different oxides of nitrogen (NO<sub>x</sub>) are measured using instrument such as Horiba and 2B technologies.

#### 3.4.2.1 CO Monitor

Horiba AP370 ambient CO monitors are used for the measurement of CO (Fig 3.9). This monitor uses the non-dispersive infrared analysis method as its principle of operation (Fig. 3.10). The instrument is capable of measuring continuously. The instrument can measure the concentration such as rolling average and momentary values. The measured data can be collected through the RS232 cable. The instrument can detect the CO concentration in the air sample having temperature from 5°C to 40°C. The flow rate of the instrument is 1.5L/min for air sample and for calibration

gas it was 2.5L/min. The response rate of the instrument is 60s. The minimum detection limit of the instrument is 0.05 ppm, zero drift varies from  $\pm 0.1$  ppm/24hr. The linearity of the instrument is  $\pm 1.0\%$  with a minimum time scale of 3 minutes.



Figure 3.9 Picture of the Horiba AP370 CO monitor

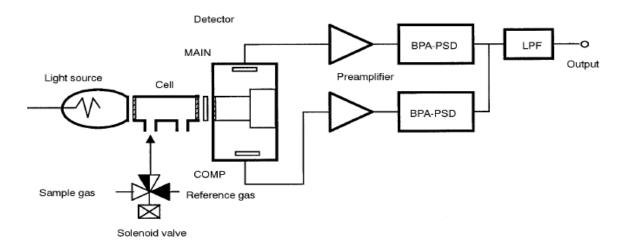


Figure 3.10 Circuit diagram of the instrument Horiba AP370 CO monitor

#### 3.4.2.2 SO<sub>2</sub> Monitor

Horiba AP370 ambient SO<sub>2</sub> monitors are used for the measurement of SO<sub>2</sub> (Fig 3.11). This monitor uses the ultraviolet fluorescent and non-dispersive infrared analysis method as its principle of operation (Fig 3.12). The instrument is capable of measuring

continuously. The instrument can measure the concentration such as rolling average and momentary values. The measured data can be collected through the RS232 cable. The instrument can detect the SO<sub>2</sub> concentration in the air sample having temperature from 5°C to 40°C. The flow rate of the instrument is 1.5L/min for air sample and for calibration gas it was 2.5L/min. The response rate of the instrument is 60s. The minimum detection limit of the instrument is 0.05 ppb, zero drift varies from  $\pm 0.1$ ppm/24hr. The linearity of the instrument is  $\pm 1.0\%$  with a minimum time scale of 5-minutes.



Figure 3.11 Picture of the Horiba AP370 SO<sub>2</sub> monitor

During measurement when the sampled air is irradiated with UV ray (215nm), the SO<sub>2</sub> pollutant emits light of different wavelengths such as 320, 240 and 420nm) from the irradiated light. The irradiated light is referred as excitation light and the emitted light is referred as fluorescence. This method of absorbing the fluorescence intensity is known as fluorescence method. The fluorescence is usually detected at right angles to prevent interface from the excitation light which emits in all the direction.

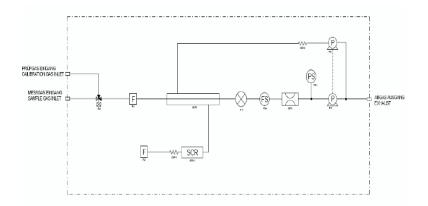


Figure 3.12 Circuit diagram of the instrument Horiba AP370 SO<sub>2</sub> monitor

#### 3.4.2.3 NO/NO<sub>2</sub>/NO<sub>x</sub> Monitor

The model 405nm NO<sub>x</sub> monitor is an instrument used for the direct measurement of atmospheric nitrogen dioxide, nitric oxide and as well as the total reactive oxides of nitrogen (NO<sub>x</sub> = NO+NO<sub>2</sub>) (Fig 3.13). The concentration ranges from 0 to 10 ppm for NO<sub>2</sub> and 0 to 2 ppb for NO (Fig. 3.14). The instrument can measure the pollutants with high precision and accuracy. The NO<sub>2</sub> measurement is based on the direct absorbance at 405 nm. NO is measured by selective conversion with ~ 100% efficiency with a highly selective reaction of NO with Ozone (O<sub>3</sub>). Total NO<sub>x</sub> is obtained by addition of NO and NO<sub>2</sub>. The long-life pump of the light source is ~15,000 hours with an internal data logger. The operating temperature of the instrument is 10° to 50° C. The minimum flow rate of the instrument is 1.4 L/Min with a maximum flow rate of 1.6 L/min. The data can be obtained for 5s, 1min, 5min and 1-hour interval.



Figure 3.13 Picture of the NO<sub>2</sub>/NO/NO<sub>x</sub> monitor

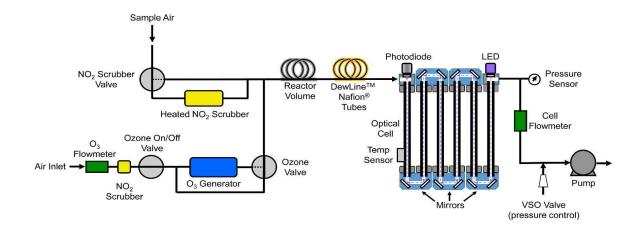


Figure 3.14 Circuit diagram of the NO<sub>2</sub>/NO/NO<sub>x</sub> monitor

## 3.4.2.4 Ozone Monitor



Figure 3.15 Picture of the 2B Technologies O<sub>3</sub> monitor

The 205 2B technologies  $O_3$  monitor makes use of dual beam technology (Fig 3.15). In this dual beam technology, the UV light measurements are made simultaneously with both ozone-scrubbed air and un scrubbed air. The instrument uses UV based  $O_3$ measurements are one of the fastest measurement techniques available in the market. The operating temperature of the instrument is 10° to 50° C (Fig 3.16). The Nominal flow rate of the instrument is 1.8 L/Min. The data can be obtained for 10s, 1min, 5min and 1-hour interval.

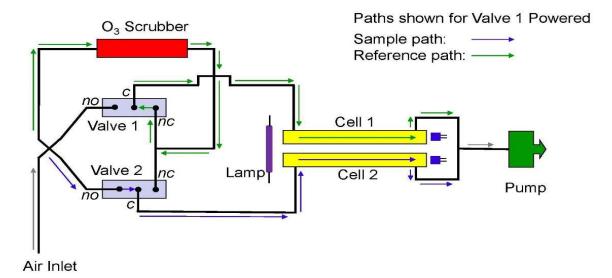


Figure 3.16 Circuit diagram of the 2B Technologies O<sub>3</sub> monitor.

# 3.4.3 Meteorological measurement

The meteorological parameters such as wind speed, wind direction, relative humidity and temperature are measured using the Amil make sensor which has an accuracy of  $\pm$  5% and sensitivity 0.5% (Fig 3.17). The weather station is fully automated and real-time instrument.



Figure 3.17 Picture of the automated weather station

# 3.4.4 Secondary data collection

The secondary data's such as solar radiation, planetary boundary layer is taken from the open source data sets. The solar radiation is taken from the central ambient air quality monitoring station which is located at an aerial distance less than 500 m from the study region. Similarly, the planetary boundary layer height for the study region during the different season is taken from the ERA5 reanalysis data with an interval of 1-hr with a resolution of  $0.25^{\circ} \times 0.25^{\circ}$ . The both primary and secondary data are taken from well validated sources for minimal errors and used in the study.

# 3.5 Data analysis and visualization

Data analysis is done using different software such as R Software, Origin software, MATLAB software and the health impact assessment are done using the mathematical and computational models.

# 3.5.1 R software

R software (version 5.12.8) is one of the widely used programming language used for statistical analysis and data visualization (Fig. 3.18). This is an open source software which is freely available under general public license. The software has excellent

features in data mining, calculation, graphical display of the data, handling of multiple data sets parallelly. This software has a wide range of libraries in which different libraries are used for various different functions. The most widely used library for the air pollution data analysis is the open-air package. This software is available for different operating system and it has a good community support. The software has different types of resources available for the learning and troubleshooting the issues. This salient feature makes it a versatile software for usage (https://cran.r-project.org/doc/contrib/Paradis-rdebuts\_en.pdf).

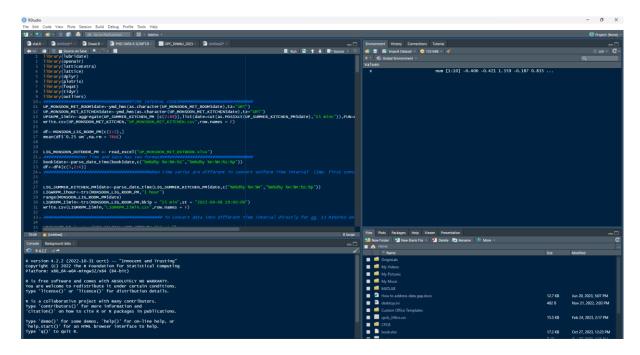


Figure 3.18 Picture of R software platform

# 3.5.1.1 Open air package

Open air package was developed by Carslaw D.C. and K. Ropkins for analysing the different air pollutants for multiple data (Fig 3.19). The open-air package provides sophisticated analysing tools such as wind rose, pollution rose. The advance tools such as bivariate polar plots and condition probability functions for identifying the different pollution sources. The package provides access to the NOAA trajectories as well for the analysis. The detailed manual for the open-air package are found in (https://davidcarslaw.com/files/openairmanual.pdf).

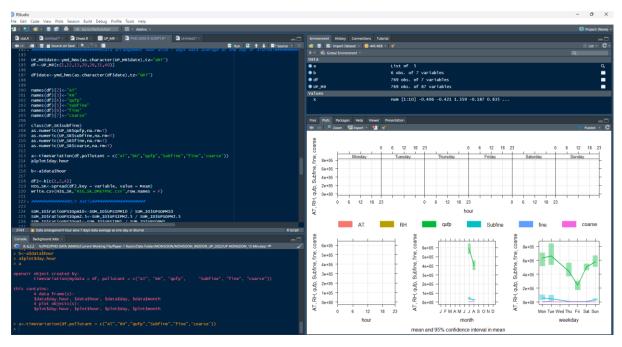


Figure 3.19 Data visualization in R platform

# 3.5.2 MATLAB software

MATLAB is a tool used to analyses, visualize and handle different types of air quality data (Fig 3.20). The software can be used for creation of plots, dashboards etc (Fig 3.21).

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*Figure 3.20 Picture of MATLAB software with data feeding window* 



Figure 3.21 MATLAB data outputs and visualization

The timeseries data analysis, heatmaps can be created using the air quality data in MATLAB. MATLAB can be integrated with IOT platforms to analyze data real time using the sensors. The software can be used for prediction of air quality using machine learning tools. Regression analysis can also be performed using the MATLAB. This software is used for data analysis and visualization in the study. The version used in the study is 2016a.

#### 3.5.3 ORIGIN software

Origin is a data analysis and graphing software which was developed by Origin Lab corporation (Fig 3.22). Origin software is helpful in plotting 2D and 3D plots. This software is useful in data analysis such as statistical analysis, signal processing and curve fitting. Curve fitting uses Levenberg-Marquard algorithm in this software. In origin programming languages such as python and R functions are enabled and also scripting language like Lab Talk. In origin different data formats such as Net CDF, excel and ASCII text formats. The output of the origin can be taken in formats such as JPEG, PNG and TIFF files. This study used Origin 2022 version for analysis (Fig 3.23).

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2	15342.15	767.1074	1534.215	27842.97	1392.149	21381.57577	1069.079	17601.09		10531.03	526,5516		0.01118	5.589E-4	0.02029	0.00101	0.01558	7.789E-4	0.01282	6.412E-4	0.00767	3.836E-4			
	15369.02	768.4510	1536.902	26780.48	1339.024	20722 22839	1035.111	17714.02		10768.64	538.4320		0.01458	7.2887E-4	0.0254	0.00127	0.01965	9.827E-4	0.0168	8.401E-4	0.01021				
	15725.37	786.2686	1572.537	26545.17 26228.11	1327.258	20494 94565 20285 33588	1024.747	18365.52		11257.94 11739.36	585,9582		0.01942	9.712E-4 0.00129	0.03279	0.00164	0.02532	0.00127	0.02269	0.00113	0.01391	6.953E-4 9.4452E-4			
	16221.75		1622.175	26107.58		20206.53566		19606 15	980.3077	12405.12	620.2562		0.02571	0.00129	0.05476	0.00274	0.03265	0.00163	0.04113	0.00193	0.02602	0.0013			
1	16509.27		1650.927	26081.80		20155.34539		20241.47			649.6121		0.04516	0.00226	0.07135	0.00357	0.05514		0.05537	0.00277	0.03554	0.00178			
	16943.82	847.1909	1694.382	26148.21	1307.410	20223 37096	1011.169	20929.09	1046.455	13652.20	682.6100		0.05048	0.00302	0.09333	0.00467	0.07218	0.00361	0.0747	0.00374	0.04873	0.00244			
	17311.35	865.5675	1731.135	25931.49		20125.37568		21312.96		14255.03	712.7516		0.08067	0.00403	0.12085	0.00604	0.09379	0.00469	0.09932	0.00497	0.06643	0.00332			
1	17942.19		1794.219 1874.613	25801.08 25682.69	1290.054 1284.134	20178.74576 20222 33942		21698.47 22048.51	1084.924 1102.426	14940.98 15693.78	747.0491 784.6892		0.10924	0.00546	0.15709	0.00785	0.12286	0.00614	0.13211 0.17553	0.00661 0.00878	0.09097	0.00455			
11			1974.963	25508.23		20142.2799	1007.114	22126.51	1106.325	16505.95	825.2973		0.20574	0.01029	0.26573	0.01329	0.20983	0.01049	0.2305	0.01153	0.17195	0.0086			
13			2112.503	25337.86		20092.18058	1004.609	22013.99		17410.27	870.5134		0.28826	0.01441	0.34574	0.01729	0.27416	0.01371	0.30038	0.01502	0.23757	0.01188			
14			2265.363	25222.45		19981.0428		21788.20			915.9933		0.40531	0.02027	0.45127	0.02256	0.35749		0.38982	0.01949	0.32777	0.01639			
1!	24492.01		2449.201 2661.007	25072.20	1253.61	19857.40103 19777.63112	992,8701	21416.87 21041.83	1070.843	19265.85	963.3424		0.57523	0.02876	0.58886	0.02944	0.46538	0.02332	0.503	0.02515	0.45251	0.02263			
1	28922.98		2892.298	25037.45		19679 14411		20696.67	1034.833	21515.79	1075 790		1.17531	0.05877	1.01741	0.05087	0.79967	0.03998	0.84102	0.04205	0.87431	0.04372			
11	31406.65	1570.333	3140.665	25072.84	1253.642	19659 29698	982.9648	20292.67	1014.633	22784.31	1139.215		1.68264	0.08413	1.3433	0.06717	1.05326	0.05266	1.0872	0.05436	1.22069	0.06103			
11			3386.988	25166.58		19631.74299		19936.99	995.8494	24088.30	1204.415		2.3968	0.11984	1.7809	0.08905	1.38923	0.06946	1.41084	0.07054	1.7046	0.08523			
20			3616.377 3815.008	25321.38 25513.42	1266.069	19653.75515 19651.38932	982.6878	19572.28 19122.39	978.6138 956.1197	25316.95 26505.91	1265.847		3.38702 4.73974	0.16935	2.37156 3.16977	0.11858	1.84073 2.44147	0.09204	1.8331	0.09165	2.37114 3.29306	0.11856			
2			3964.062	25616.41	1280.820	19635.16076		18586.61		27613.17	1380.658		6.54967	0.32748	4.2325	0.21163	3.24441	0.16222	3.071	0.15355	4.56242				
2	40564.12		4056.412	25695.61	1284.781	19569.29887	978.4649	17928.20		28630.58	1431.529		8.93887	0.44694	5.6624	0.28312	4.31235	0.21562	3.9507	0.19754	6.30916	0.31546			
24	40863.55		4086.355	25605.79		19420.61422		17134.44	856.7222	29489.98	1474.499		12.04836	0.60242	7.54976	0.37749	5.72607	0.2863	5.052	0.2526	8.69497	0.43475			
2	40538.91	2026.846	4053.691 3970.375	25322.02 24851.75	1266.101 1242.587	19121.73872 18603.51443		16244.88 15156.40	812.2442 757.8199	30261.45	1513.072	100	16.04954 21.19402	0.80248	10.02558	0.50128	7.57073	0.37854	6.43171 8.09054	0.32159	11.98127 16.48693	0.59906			
21	39703.75	1985.187	3970.375	24851.75	1242.587	17801.41216		13957 78	697.8891	30885.67	1544.284	100	27.81689	1.0597	17.43069	0.87153	12.86977	0.64349	10.09096	0.40453	22 59 192	1 1296			
21	36784.50	1839.225	3678,450	23040.27	1152.014	16717.72555	835.8863	12612.38		31214.89	1560.745		36,2005	1.81002	22.67447	1.13372	16.45229	0.82261	12.41213	0.62061	30,71931	1.53597			
21	34699.26	1734.963	3469.926	21678.38	1083.919	15373.85691		11180.19		30619.76	1530.988		46.74767	2.33738	29.20555	1.46028	20.71197	1.0356	15.0622		41.25147	2.06257			
31	32157.46			19958.90	997.9451 896.5945	13777.4404		9704.889	485.2444	29358.32	1467.916		59.68168	2.98408	37.04221 46.16707	1.85211 2 30835	25.56979	1.27849	18.01153	0.90058	54.4866	2.72433			
3	29233.58	1461.679	2923.358 2503.477	1/931.89		12063.41037 10291.80942		8278.855	413.9433 346.7869	27437.33 25028.96	13/1.867		75.26411 93.7005	3.76321	46.16707	2.30835	31.05814 37.04073		21.31459 24.96215	1.065/3	70.63958 90.08074	4.50404			
33	22521.09		2252.109	13317.52		8513.86682		5731.347	286.5674	22208.58	1110.429		114,2626	5.71313	67.56772	3.37839	43.19586		29.07855	1.45393	112.6772	5.63386			
34	19002.28	950.1139	1900.228	10961.26		6819.2778		4675.600		19208.30	960.4149		137.1493	6.85746	79.11335	3.95567	49.21824	2.46091	33.74629	1.68731	138.6368	6.93184			
35	15420.10		1542.010	8646.087		5140.23807	257.0119	3691.381	184.5691	15983.00	799.1501		159.8724	7.99362	89.63994	4.482			38.27132	1.91357	165.7082	8.28541			
31	11973.46	598.6728 444.8713	1197.346 889.7425	6569.259 4777.831	328.4630 238.8915	3673.32675 2457.80559	183.6663 122.8903	2805.141 2007.655	140.2570 100.3828	12753.93 9615.904	637,6965 480,7952		180.1547	9.00773	98.84212 105.4405	4.94211 5.27203		2.76348	42.2067 44.30617	2.11034	191.8973 212.2099	9.59487			
3	6175.664		617 5664	3261 429		1519,2923	75.96461	1335 779		6676.295	333.8148		202.0628	10.10314	105.4405	5.33557	49.71		43.70579	2.18529	212.2055				
31	4167.926		416.7926	2245.648		966.83047	48.34152	854.1228			218.2749		204.3645	10.21823	110.1595	5.50798	47.40624		41.87965		214.0514				
41	2540.122		254.0122	1396.184		549.88791	27.4944	488.6142		2519.624	125.9812		188.6045	9.43023	103.6673	5.18337	40.82921		36.27968	1.81398	187.0835	9.35418			
4	1280.059		128.0059 81.03179	739.8211 555.0250	36.99105	255.58016	12.77901 11.92228	232.2445	9 31034	1184.177 677.2177	59.20887 33.86088		145.3703	7.26852	84.01816 97.3144	4.20091 4.86572	29.0249	1.45124 2.09036	26.37476 32.64736	1.31874	134.4809	6.72405 5.93698			
4			41.08564	327.9635		238 4450	8.03836	186.2067 261.1669	9.31034	326.3451	33.86088		142.0759	5.61027	97.3144	4.80572	41.80721 43.90392		32.64736	3.56578	118.7395	4.45621			
4		11.31865			11.13973	154.34778		65.68106	3.28405	186.9775	9.34888	1000	97.07327	4.85366	95.53935	4.77697	66.18727		28.16736	1.40837	80.18008	4.009			
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Figure 3.22 Picture of data feeding platform in the origin software

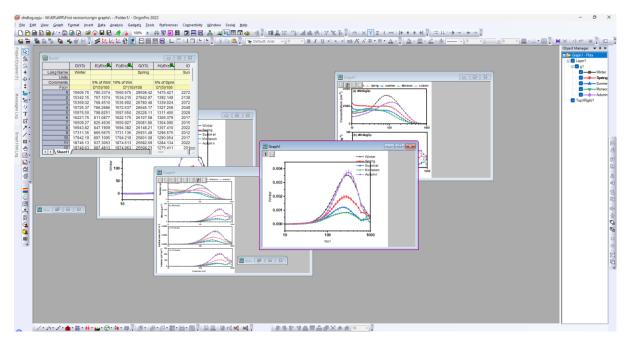


Figure 3.23 Output of the data visualization in the Origin software

# 3.5.4 IBM SPSS software.

The IBM SPSS software in the study is used for conduction different statistical tests (Fig 3.24). The air quality data is tested for its significance using the IBM SPSS

software (Fig 3.25). The Levene test was conducted for testing the variance equality during the different emission periods of the study and found that the p value is found to be < 0.05. The other statistical tests such as F test are also utilized. The variances at 95% confidence interval level are also applied in the study. The outliers in the data were statistically tested and removed using the  $z \pm 3$  score values. The Kruskal-Wallis test is used for determining the differences of the data between the different periods. The Kruskal-Wallis test is a non-parametric test. The Mann-Whitney U (non - parametric test) is used for testing the variation of independent groups especially season in our case. The statistical tests reveal that the concentration of the nanoparticles in the study area is statistically significant based on the sources and seasons. The p-value shows < 0.05 and the null hypothesis is rejected.

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1	1		January	1	I Residential		1	2018		1 Punjabi Bagh	59	
2	2	2	Febraury	1	I Residential		1	2018		1 Punjabi Bagh	60	
3	3		March		I Residential		1	2018		1 Punjabi Bagh	59	
4	4	4	April	1	I Residential		1	2018		1 Punjabi Bagh	60	
5	5	5	May		I Residential		1	2018		1 Punjabi Bagh	60	
6	6	6	June	1	I Residential		1	2018		1 Punjabi Bagh	62	
7	7		July		I Residential		1	2018		1 Punjabi Bagh	60	
8	8	8	August	1	I Residential		1	2018	-	1 Punjabi Bagh	60	
9	9		September		I Residential		1	2018		1 Punjabi Bagh	60	
10	10	10	October	1	I Residential		1	2018		1 Punjabi Bagh	58	
11	11		November	1	I Residential		1	2018		1 Punjabi Bagh	58	
12	12	12	December	1	I Residential		1	2018		1 Punjabi Bagh	58	
13	13	1	January	1	I Residential		2	2019		1 Punjabi Bagh	59	
14	14	2	Febraury	1	I Residential		2	2019		1 Punjabi Bagh	61	
15	15	3	March	1	I Residential		2	2019		1 Punjabi Bagh	59	
16	16	4	April	1	I Residential		2	2019		1 Punjabi Bagh	60	
17	17	5	May	1	Residential		2	2019		1 Punjabi Bagh	60	
18	18	6	i June	1	I Residential		2	2019		1 Punjabi Bagh	60	
19	19	7	July	1	Residential		2	2019		1 Punjabi Bagh	60	
20	20	8	August	1	I Residential		2	2019		1 Punjabi Bagh	60	
21	21	g	September	1	I Residential		2	2019		1 Punjabi Bagh	60	
22	22	10	October	1	Residential		2	2019		1 Punjabi Bagh	57	
23	23	11	November	1	I Residential		2	2019		1 Punjabi Bagh	58	
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Figure 3.24 Data feeding in SPSS for statistical analysis

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Year		Year	Leq (Day)	Leq (Night)	Ldn		
🔓 Leq (Night	N Valid	600	600	600	600		
Ldn	Missing	0	0	0	0		
Histogram	Mean	2020.00	61.50	57.21	64.73253250		
Year	Std. Error of Mean	.058	.266	.295	.2735782102		
🛗 Leq (Day)	Median	2020.00	61.00	56.00	63.27951481		
- 🚡 Leq (Night	Mode	2018 <sup>a</sup>	57	55	69.68712746		
	Std. Deviation	1.415	6.513	7.215	6.701270197		
Descriptives	Variance	2.003	42.424	52.063	44.907		
Title	Skewness	.000	.254	.841	.761		
Notes	Std. Error of Skewness	.100	.100	.100	.100		
🛍 Log	Kurtosis	-1.301	.024	.392	.412		
Explore	Std. Error of Kurtosis	.199	.199	.199	.199		
Title	Range	4	37	40	38.22548368		
Land use ID	Minimum	2018	45	44	51.02059991		
👘 Title	Maximum	2022	82	84	89.24608359		
Case Proc	Sum	1212000	36900	34323	38839.51950		
M-Estimate	Percentiles 25	2019.00	57.00	52.00	59.99657754		
- 🍎 Percentiles	50	2020.00	61.00	56.00	63.27951481	Activate Windows	
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Figure 3.25 Results of the statistical analysis in SPSS

# 3.5.5 Inhalable nanoparticle concentration.

The inhalable particle number concentration is a mathematical method of analyzing the probable inhalable concentration of the nanoparticles in the atmosphere based on the concentration of the nanoparticles and also the different breathing rate associated inhalation volume. Different physical activities of human will have different inhalation volumes which influence the exposure of the air pollution as well (Prabhu et al., 2019; Qiu et al., 2019) (Table 3.4). This mathematical method is considered as the reliable method and also one of the easiest methods for analyzing the probable inhalable particle number concentration. The analysis shows the number of particles which can be penetrated into the human respiratory systems. The nanoparticles are believed to have deposition in the deeper regions of the lungs such as alveoli and bronchioles (Kim et al., 2017; Koehler and Peters, 2015).

The equation which is used for the analysis of the inhalable particle number concentration is given below. where IPN represents the inhalable particle number concentration, PNC is particle number concentration (concentration of the nanoparticles) and the IR represents the Inhalation rate of that particular physical activity.

$$IPN(min^{-1}) = PNC(cm^{-3}) \times IR(cm^{3}min^{-1}) - (15)$$

The different IR rates used in the study is based on the guidelines of the US EPA exposure guideline handbook.

The IR values will be more for heavy physical activity such as running, weightlifting, heavy exercise. During the heavy exercise period the concentration of the particles gets deposited in the respiratory system will be 4.5 times higher than the normal mode of deposition during the resting position. (Ma et al., 2022). The walking activity is found to have less IR value and the long distance or cross-country running is found to have the higher IR which is used in our study. The detailed IR of the different physical activity is provided in the Table 3.4. The advantage of this method is that it requires very less inputs, i.e. only concentration of the nanoparticles is sufficient for estimation. This is the simplest mathematical calculation method for exposure estimation. The limitation of this method is that this method provides only the probable number concentration which can gets deposited in the respiratory system. The actual deposition in the respiratory system may vary in real which is not estimated by this method.

Category		Associated activity	Inhalation volume (cm <sup>-3</sup> /min)
Light	Level 1	Walking, cloth washing.	13
	Level 2	Scrubbing floors, bowling.	19
	Level 3	Dancing, pushing, wheelbarrow, construction activities.	25
Medium	Level 1	Easy cycling.	30

Table 3.4. Table showing different Inhalation rates and physical activity associated with it.

Le	evel 2	Climbing stairs, playing tennis.	35
Le		Fast walking and digging trenches, potholes.	40
Heavy Le	evel 1	Climbing stairs with load, playing squash and handball.	55
Le	evel 2	Wood cutting with an axe.	63
Le		Activities in level 1 and 2 simultaneously.	72
Very Heavy		Running, competitive cycling.	85
Severe		Long-distance running, cross- country skiing.	100

#### 3.5.6 Particle dosimetry model

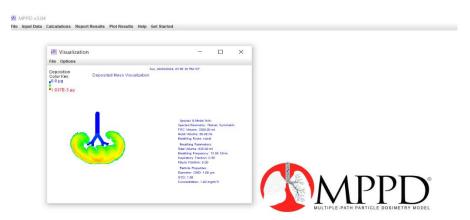
The nanoparticles deposition in the different regions of the human respiratory system can be evaluated using a highly sophisticated model called Multiple Path Particle Dosimetry model (MPPD 3.04) (Fig 3.26). The MPPD model was developed by Owen price along with the Hammer institute of health science and Applied Research Associates. In the MPPD model calculates the deposition and also the clearance of monodisperse and polydisperse particles in the respiratory tracts of humans and different animal species. The model can predict deposition of particles from 1 nm to 100  $\mu$ m. The model is also useful for toxicology studies of different laboratory animals such as rat, mouse pigs and rabbit. The model depositions prediction uses both single and multiple path flow analysis. In single path model specification, the deposition analysis of the nanoparticles is done using typical airway path generation but in multiple path method the deposition of the exposed particles is calculated using the theoretically derived efficiencies such as deposition by diffusion, sedimentation, and also the impaction within the airway. The filtration of the nanoparticles by the defensive mechanism of nose and mouth are also determined using the empirically derived efficiency equations. This model is considered as one of the most accurate models for dosimetry analysis. The model works on the basis of computational fluid dynamics. The more general details of the model can be found at (https://www.ara.com/mppd/).

This model is a well validated model and it is tested and widely used in research and education purpose for the nanoparticle deposition in the human respiratory tract (Khan et al., 2022; Manojkumar and Srimuruganandam, 2021b, 2022b). This model is considered to be one of the best models available for the dosimetry studies due to the features and updated functions of the model. The major input parameters of the model include, particles feature such as size, density and then lung features such as breathing frequency (BF), tidal volume (TV), upper respiratory tract volume (URT), and the different orientation of exposure conditions. The different exposure orientation conditions available in the model are particles recaching the respiratory system through respiratory tract through nose and the body in upright orientation and dispersion. This model also uses certain assumptions and constant as per the different global study suggestions. In the model the deposition of the nanoparticles in the different regions of the respiratory system can be analyzed such as pulmonary, tracheobronchial, and alveolar regions. The depositions in the alveoli, bronchus and bronchiole. The input of the model uses size segregated measurement for better accuracy which is one of the drawbacks of the model. Another limitation of this model is that this model uses theoretical and computational functions for the deposition analysis but in real the deposition may vary in the respiratory tract. The variation of deposition is due to the water vapor present in the respiratory tract which causes particle to grow its size in the respiratory tract due to the humid conditions.

2 Exposure Scenario		×	
Acceleration of Gravity	901.0	cm/s <sup>2</sup>	
Body Orientation	Upright -		
Body Orientation: o		•	
Body Creentations (F		1.	
Body Orientation: y		•	
Aerosol Concentration	1.0	mg/m*	
Breathing Insquarery		per mentle	
Tidal Volume	625.0	mi	
Inspiratory Traction	15		
Pause fraction	0.0		
Breathing Scenario	Nasal	-	

# Figure 3.26 Data feeding in the MPPD model

The model contains the ideal lung geometries and lung assumptions which mimics the exact function and structure of the human lungs. This model has lung geometry for different age groups of humans such as infants, children and adults. The three different regions of the human respiratory tract such as head/throat, tracheobronchial and alveolar/ pulmonary. While estimating the lung deposition the density of the particle inhaled, mono or poly dispersed particles, exposure time, breathing routes are some of the major basic model requirements which has to be provided as input parameters. In this study the advanced version of the MPPD model (3.04 version) is used. This version has several advantages compared to the other previous version (Miller et al., 2016). This version of the model can calculate the both single and multiple flow deposition. This model geometry resembles the most realistic human lung structure and this can handle different distributions of the particles at a time (Fig. 3.17).

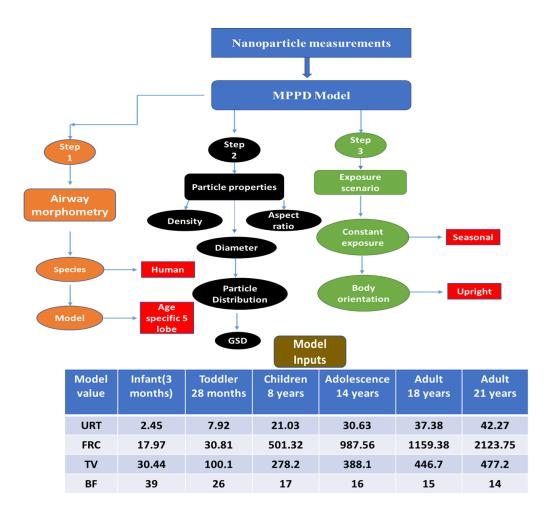


#### Figure 3.27 Demo output in the MPPD model

In the study the nanoparticles measurements are done extensively for a year. The deposition occurred during the different seasons are analysed. The seasonal mean concentration of the nanoparticles is used for the estimation deposition in the different regions of the human lungs. The deposition dose of the particles inhaled for different size ranges are analysed using the equation given below.

Dose Rate = 
$$\int_{Dp1}^{Dp2} V_E . DF(D_P) . n_N^0(D_P) . fdlog(D_P) - (16)$$

The step by step model execution is provided in the Figure 3.28 and the basic assumptions and functions of the model are also provided in the schematic diagram.



# Figure 3.28 Schematic diagram of the MPPD model

\*URT- Upper respiratory tract volume, FRC- Functional residual capacity, TV- Tidal volume, BF- Breathing frequency.

# 3.6 Summary

The detailed methodology of the study provided different information about the study area, types of instrument used and different types of software's used for data analysis and data interpretation.

# CHAPTER – 4

# **RESULTS AND DISCUSSION**

#### 4.1 Introduction

The measured atmospheric nanoparticles are classified into four different categories based on the size fractions. The classifications are  $N_{nuc}$  (10 to 30 nm), small Aitken Nsatk (30 to 50 nm), large Aitken Nlatk (50 to100 nm), Nacc accumulation mode (100 to 1000 nm). The overall size fractions (10 to 1000 nm) is referred to as N<sub>total</sub> PNC. The small Aitken and large Aitken can be combinedly referred as simply Aitken mode and the size fractions starting from the 10 to 100 nm are globally referred as Ultrafine particles (UFP 10 to 100nm). The geometric mean diameter for the entire size range is referred as GMD. The aim of the study is to measure and quantify the nanosized particle concentration in the road side microenvironment. The measurements are done in the road side environment in different emission conditions. The different emission scenarios are classified into different phases to quantify the role of transport emissions in the urban roadside regions. Similarly, the study aims to analyse the seasonal factors influence on these atmospheric nanoparticle's concentration along with the role of meteorological conditions. The measurements are done for a yearlong period in the study region which covers all the seasonal classifications in the study area. The study period includes winter, spring, summer, monsoon and autumn. Along with this the meteorological parameters such as temperature, humidity, wind direction wind speed is also measured in the study region for analysing the role of the meteorological parameters in the atmospheric nanoparticle's concentration. The dynamics of the atmospheric nanoparticles in the road side based on the emission sources variation that are classified on the basis of restrictions which was imposed to tackle the spread of Corona Virus (COVID -19). And also, the extreme pollution event in the study area was selected during the Diwali period. During this along with regular emissions the emissions from the fireworks also contributed to for higher number of atmospheric nanoparticles emissions in the study area. These scenarios were classified in to two periods namely, period I and period II. Both the period has three different phases one is before the event, another one is during the event and one is after the event. This

gives a detailed information about the role of these emissions sources in the study area. The first period is classified into three phases namely, Before Restriction (BR). During restriction (DR) and After Restriction (AR). The BR phase lasts for 20 days from 1st to 20<sup>th</sup> April 2021 in the study area, followed by DR phase which lasts for 47 days from 21<sup>st</sup> April to 5<sup>th</sup> June and the AR phase lasts for 24 days (6<sup>th</sup> June to 30<sup>th</sup> June). Totally period I covers 91 days of data measurements. Similarly, in period II the BD phase last for 33 days (3rd October to 4th November), AD period lasts for 8 days (5th to 12<sup>th</sup> November and DRII phase lasts for 18 days (13<sup>th</sup> to 30<sup>th</sup> November). Period II covers a monitoring period of 59 days. In total both the period I and period II cover a total monitoring period of 150 days. The seasonal classification in the study area includes winter, spring, summer, monsoon and autumn. The winter seasons starts from the month of December to the mid of February accounting for 76 day out of which 96 days of data is available for analysis. Similarly, for spring from mid-February to March accounting for 45 days (44 days data available). The summer seasons starts from April and ends at June which lasts for 91 days. The summer season is the longest season in the study region and during this season 84 days of data is available for analysis. The monsoon season starts in the month of July to mid-September and 41 days of data is available out of 77 days. Autumn season starts from mid-September to November consisting of 76 days and 60 days of data are available during this season. In a year out of 365 days the measurement campaign lasts for around 300 days of data in all the different periods. The study used the measured concentration for analysing the exposure to these nanoparticles using different mathematical and computerized models.

# 4.2 Analysis of particle dynamics based on the emission sources variation

#### 4.2.1 Temporal variation PNC during different periods

The particle number concentration and associated GMD concentration during the different periods are analysed using the temporal variation of the concentration. During the both the periods the lowest concentration of the total PNC ( $1.7 \times 10^3 \text{ cm}^{-3}$ ) was found during the DR phase in period I in which the vehicular movement in the study region reduced to around 45%. The study region is in the roadside microenvironment

where the majority of the emission sources are from the transportation sectors. The engine exhaust emissions release wide range of pollutants. Comparted to the normal scenario in the BR phase the concentration of the particles reduced to around 31% (Fig 4.1). This shows the role of vehicular sources in the nanoparticle's emission in the study region. The study region represents the urban roadside micro environment which represents similar conditions the road microenvironments in the city. During the AR phase when the vehicular activities resumed, the particle number concentration also starts to increase and reached peak concentration of 1.3 x  $10^5$  cm<sup>-3</sup> (Fig 4.1). This concentration is the highest recorded hourly mean concentration during the period I. During the BR and AR phase, there was no restriction for vehicular sources. The period I analysis shows that the reduction in vehicular sources directly proportional to the number concentration.

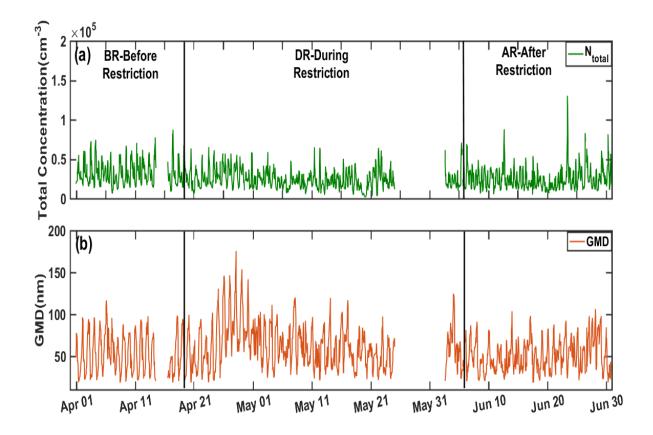
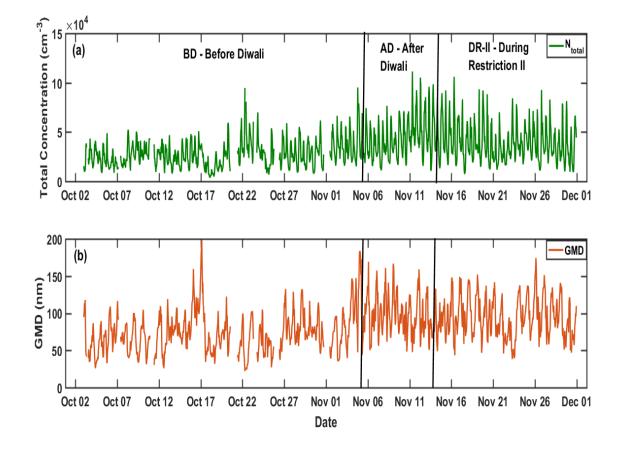


Figure 4.1: Temporal distribution of hourly average concentration of (a) $N_{total}$  (b) GMD during Period I (1<sup>st</sup> April 2021 – 30<sup>th</sup> June 2021)

During period II (Fig 4.2) the highest mean concentration of the total PNC is found during the AD phase with a concentration of about  $4.2 \times 10^4$  cm<sup>-3</sup> which is around 35% increase in concentration compared to the BD phase (2.7 x  $10^4$  cm<sup>-3</sup>). However, the concentration during the AD phase decreases gradually compared to the previous year (2.19 x  $10^6$  cm<sup>-3</sup> in 2021) due to the implementation on restrictions to the cracker's usage (Yadav et al., 2022). The concentration received during the AD phase is the highest found concentration in the both periods. The period I and II shows that the concentration of the atmospheric nanoparticles in the urban atmosphere id directly proportional to the emission sources. Higher emission sources lead to higher concentration of the particles (AD phase, period II) and lower sources leads to low emissions (DR phase, period I). The concentration received during the AD phase is 25% higher than the period I maximum concentration and 57% higher than the period I lower concentration. The concentration of the atmospheric nano particles in the study area are also driven by the local meteorological conditions and the precipitation process in the region.



# Figure 4.2: Temporal distribution of hourly average concentration of (a) $N_{total}$ (b) GMD during Period II (3<sup>rd</sup> October 2021 – 30<sup>th</sup> November 2021)

The gaps in the figures represents the data missing. In the study region the during period I and II rainfall of 116 and 80mm received respectively. The precipitation received during this phase helps in the wet removal of the particles. Global studies found that the nanoparticle concentration in the study region varies from  $10^2$  to  $10^6$  particles cm<sup>-3</sup> (Kumar et al., 2014). In our study the concentration varies from  $10^3$  to  $10^5$  particles cm<sup>-3</sup>. During different periods the order of concentration of the total PNC varies as BR> AR> DR phase in Period I and DR II>AD>BD in period II (Table 4.1). The minimum concentration of the atmospheric nanoparticles increases in urban regions in the recent years especially in the south Asian region due to the intense anthropogenic emission activities in these regions and many globally polluted urban regions are found in these Global south regions which a serious concern (Ramachandran and Rupakheti, 2022).

Phase	Size (cm <sup>-3</sup> )	Minimum	Maximum	Mean $\pm$ S. D	Median
Before Restriction (BR)	N nuc	534	74375	14054 ±13242	10317
	N satk	905	19581	4768 ±2791	4127
	N latk	1654	18248	6256 ±3189	5830
	N acc	705	20206	6521 ± 3952	6040
	N total	6362	87343	$31599 \pm 15028$	28220
	GMD (nm)	18	117	52 ± 23	49
During Restriction	N nuc	106	54471	8131 ± 8691	111
(DR)	N satk	305	19245	$3399 \pm 2393$	12
	N latk	498	19316	$5604 \pm 3153$	5283
	N acc	418	33838	$7346 \pm 5946$	2801
	N total	1752	70927	$24481 \pm 12034$	4766
	GMD (nm)	18	176	77 ± 16	7
After Restriction (AR)	N nuc	565	90646	$10816 \pm 10927$	7537

Table 4.1: Statistical summary of the different size range concentration during different periods

N satk	727	25546	$4123 \pm 2991$	3262
N latk	966	19861	$5382\pm2901$	4700
N acc	709	15093	$4912\pm2541$	4424
N total	5022	130433	$25235 \pm 14173$	22034
GMD (nm)	18	106	50 ± 18	47
N nuc	161	71873	$6917\pm7070$	4741
N satk	498	14646	3822 ± 2513	3252
N latk	897	21177	$6447\pm3893$	5196
N acc	1012	58056	$10236\pm 6460$	9086
N total	3774	95021	$27423 \pm 13498$	2554
GMD (nm)	22	198	73 ± 26	70
N nuc	504	46986	$7823 \pm 7713$	5273
N satk	530	14008	$4343 \pm 3145$	3590
N latk	2091	28951	$8947\pm6527$	6858
N acc	6513	45886	$21315\pm7874$	21016
N total	11256	111316	$42429\pm19320$	39260
GMD (nm)	44	184	$100 \pm 28$	97
N nuc	204	37291	$5862 \pm 6363$	295
N satk	479	17876	$4384 \pm 3373$	655
N latk	1565	36062	$10197 \pm 7144$	1117
N acc	3813	48752	$19065 \pm 8909$	885
N total	7505	105887	39510 ± 19717	35039
GMD (nm)	38	174	94 ± 27	91
	N latk N acc N total GMD (nm) N nuc N satk N latk N acc N total GMD (nm) N nuc N satk N latk N acc N total GMD (nm) N nuc N satk N latk N acc N total GMD (nm) N nuc N satk N latk N acc N total N satk N latk N acc N satk N latk N acc N satk N latk N satk N latk N satk N latk N satk	N latk         966           N acc         709           N total         5022           GMD (nm)         18           N nuc         161           N satk         498           N latk         897           N acc         1012           N total         3774           GMD (nm)         22           N total         3774           GMD (nm)         22           N nuc         504           N satk         530           N latk         2091           N satk         530           N latk         2091           N acc         6513           N total         11256           GMD (nm)         44           N nuc         204           N satk         479           N latk         1565           N acc         3813           N total         7505	N latk         966         19861           N acc         709         15093           N total         5022         130433           GMD (nm)         18         106           N nuc         161         71873           N satk         498         14646           N latk         897         21177           N acc         1012         58056           N total         3774         95021           GMD (nm)         22         198           N nuc         504         46986           N satk         530         14008           N latk         2091         28951           N acc         6513         45886           N total         11256         111316           GMD (nm)         44         184           N nuc         204         37291           N satk         479         17876           N latk         1565         36062           N acc         3813         48752           N total         7505         105887	Nlatk96619861 $5382 \pm 2901$ Nacc70915093 $4912 \pm 2541$ Ntotal5022130433 $25235 \pm 14173$ GMD (nm)18106 $50 \pm 18$ Nnuc16171873 $6917 \pm 7070$ Nsatk49814646 $3822 \pm 2513$ Nlatk89721177 $6447 \pm 3893$ Nacc10125805610236 \pm 6460Ntotal377495021 $27423 \pm 13498$ GMD (nm)22198 $73 \pm 26$ Nnuc50446986 $7823 \pm 7713$ Nsatk530140084343 \pm 3145Nlatk209128951 $8947 \pm 6527$ Nacc65134588621315 $\pm 7874$ Ntotal11256111316 $42429 \pm 19320$ GMD (nm)44184100 $\pm 28$ Nnuc204 $37291$ $5862 \pm 6363$ Nsatk47917876 $4384 \pm 3373$ Nlatk1565 $36062$ 10197 $\pm 7144$ Nacc3813 $48752$ 19065 $\pm 8909$ Ntotal7505105887 $39510 \pm 19717$

# 4.2.2 Geometric Mean diameter analysis in different periods

The geometric mean diameter of the particle's concentration based on the types of sources. The concentration of the GMD shows more deviation when the sources are complex and it shows very less deviation when the source present in the region is not complex possibly from one or two sources with similar pattern of the emission.

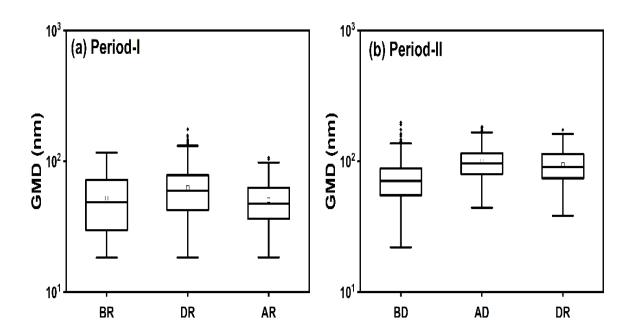


Figure 4.3: Box whisker plot for an hourly average concentration of GMD during Period I (1<sup>st</sup> April 2021 to 30<sup>th</sup> June 2021) and Period II (3<sup>rd</sup> October 2021 to 30<sup>th</sup> November 2021)

The GMD represents the emission source of the particle. The literature suggests the different size range of the particles are based on the fuels. The CNG fuelled vehicles emits particles in the size ranged from 20 to 60 nm and the diesel-powered vehicles can emits particles in the range of 20 to 130 nm. The natural new particle formation process shows the GMD of particles < 30nm. The reduction in vehicular sources emission in period I can be clearly seen from the size range of the GMD particles measure during that period (Fig 4.3, Table 4.2). During the period I of the study, the size range of the GMD ranges from 15 to 200 nm in all the three different phases. During the restriction phase (DR phase) the movement of HCV (heavy commercial vehicles) is found more compared to the LCV (light commercial vehicles). The vehicle density of the study area is more dominated by the LCV but during the DR phase due to the restriction imposed the personnel cars (LCV) movement is observed less and the movement of HCV (goods carriers) are found more. The LCV utilizes CNG as fuel which emits particles in the size range of 20 to 60 nm and the diesel-powered HCV emits particles in higher size ranges. Apart from the source, the existing meteorological conditions also plays a major role in size of the particles as it undergoes transformation such as coagulation and condensation in the atmosphere. During the period I the GMD was less than 100nm shows that the sources in the study region is probably vehicles. In the study region the hourly peak vehicular movement was around 1300 vehicles includes all the major types of vehicles. In a day around 35000 vehicles cross the study region. During period II the GMD was in the range of  $\geq$  100 nm. The study period GMD was the highest mean GMD recorded during the study period. This was due to the presence of different sources such as Diwali Sources (crackers) apart from the vehicular sources. During the period II, the deviation of the GMD of 100 nm was less and it was higher during the period II. In period II, the GMD of 100 nm was obtained during the AD phase and on DRII phase, the GMD ranged from 75 to 100 nm. The percentage change of the vehicular density in the different zones of the study region is obtained from the google mobility data. The data is used to identify how the vehicle fleet in the study region varies during the different periods of the study.

Phase	Retail	Local	Local	Transit	Work	Residential
	Centre	markets	Parks	stations	places	areas
BR	-38.6	4.9	-21.6	-16.9	-29.6	11.6
DR	-71.4	-34.5	-56.3	-60.4	-66.7	25.9
AR	-34.0	10.2	-24.4	-20.7	-34.9	11.1
BD	-11.3	39.8	-10.2	-0.9	-15.5	6.8
AD	-12.6	34.1	-10.2	-3.4	-25.7	7.4
DR II	-11.2	33.5	-9.3	1.8	-16.3	5.2

Table 4.2: Percentage reduction of vehicles in different zones of the study area

The google mobility data reveals that the restriction imposed for the vehicular movement in the study region results in the reduction of vehicular fleet in the public areas. The residential zones in the study region sees an increase in the vehicular movement due to the gathering/flocculation of the people in their residence itself. The DR Phase in the study regions shows a maximum vehicle fleet reduction. In DR phase the maximum vehicular fleet reduction was seen in the retail centres (-71.4) followed

by work places (-66.7), transit stations (-60.4), local parks (-56.3) and local markets (-34.5%). The least reduction was seen in DRII phase (period II) in the transit stations.

# 4.2.3 Particle size distribution of the different size particles during different periods

The size distribution of the entire particle size ranges during different periods also signifies the size profiling of the atmospheric nanoparticles. The ultrafine particles are found to have more association with the engine exhaust emission in the urban regions and about 80 to 90% of the total PNC emission can be from the UFP size ranges (Rönkkö and Timonen, 2019). During period I, the vehicular sources are found less especially during the DR phase which resulted in less particle number size distribution. During the AD phase when the Diwali emissions are found the concentration and size distribution of the particles increased and peak concentrations was found in the accumulation mode. This shows that the different emissions sources such as engine exhausts and Diwali fireworks emits particles in the different size ranges namely ultrafine and accumulation mode particles. During the AD phase the study region climate falls under the winter season where the coagulation of the emissions was found more (Sabaliauskas, et al., 2012; Schneider et al., 2015). This phenomenon also contributed to a certain extent for the higher accumulation size particles during the AD phase. The size distribution profiles of the size ranges 10 to 1000 nm during the different periods and phases of the monitoring period is shown in the Figure (4.4).

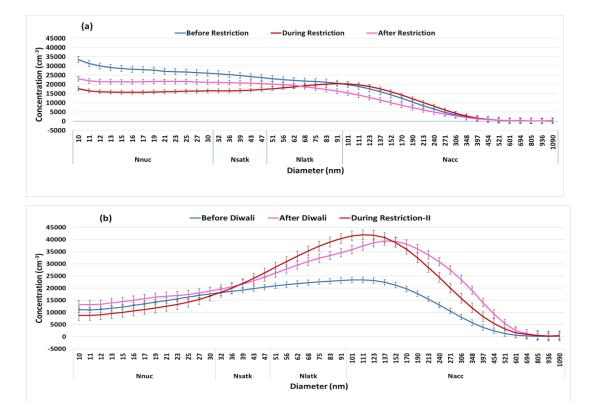


Figure 4.4: Particle number size distribution of hourly average size concentration ranges from 10 to 100 nm during Period I ( $1^{st}$  April 2021 –  $30^{th}$  June 2021) to Period II ( $3^{rd}$  October 2021 –  $30^{th}$  November 2021)

The size distribution clearly shows that the concentration of the lower size particles (< 100 nm) is high during the period I where the climate is hot and less humid. The particles concentration gradually starts to reduce from higher size to lower size. In period II the complex sources lead to the lesser concentration of the UFP particles then flowed by higher concentration of accumulation mode (100 to 500 only) and gradual reduction of the particles having diameter > 500nm. The maximum concentration of the particles during the period I is ~  $3.5 \times 10^3$  cm<sup>-3</sup> which is recorded for the particle having diameter ~10 nm. Similarly, in the period II the concentration of the particles around 100 to 200 nm diameter reaches a peak concentration of about  $4.5 \times 10^3$  cm<sup>-3</sup>. The order of concentration from higher to lower for particles till 10 nm in the period I is BR > AR > DR and in period to for the particles having diameter 100 to 200nm the order is DR II> AD > BD.

# 4.2.4 Time series measurements of meteorological parameters during different periods

During period I and period II the meteorological parameters such as temperature, relative humidity, wind speed and solar radiation is also measured along with the nanoparticle's concentration. The temperature and humidity play a role in condensation of the particles whereas wind speed and direction play a role in ventilation or dispersion of the particles (Kozawa et al., 2012b; Mehel and Murzyn, 2015). During the period I the seasonal condition in the study region is summer where the temperature ranges from 30 to 36 °C and relative humidity ranges from 30 to 55%. During this period the PNC concentration shows that the particles under 100 nm are high due to emission as well as due to the unfavourable condition for coagulation of the particles. During this period, the wind speed was also >2m/s which is favourable for the particle's dispersion. During period I (Fig 4.5, Table 4.3) along with emission source reduction in the emission sources the climatology of the study area also played a major role in the concentration of the atmospheric nanoparticles. During period II (Fig 4.6, Table 4.3) the concentration of the nanoparticles was found to have more due to the complex emissions and also the particles are found in the accumulation mode range. During that period the relative humidity in the study region varies from 60% to 80% and also the less favourable wind speed (< 2m) for dispersion. The seasonal factors such as winter and summer climatic conditions also catalyzed the atmospheric nanoparticles concentration of a region apart from the intensity of the sources (Hagler et al., 2012). The solar radiation received during the period I was more when compared to the period II. It is evident from the study that the nanoparticle concentration of a region should be measured along with the regional meteorological conditions. This provides a detailed information about the dynamics of the atmospheric nanoparticles in that particular study area. The role of the meteorological conditions is discussed in detail in the later part of the thesis.

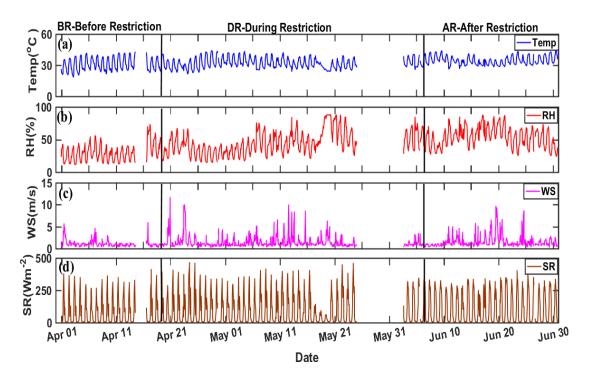


Figure 4.5: Temporal distribution of hourly average concentration of different parameters (a) Temp, (b) RH, (c) WS & (d) SR during Period I ( $1^{st}$  April 2021 –  $30^{th}$  June 2021)

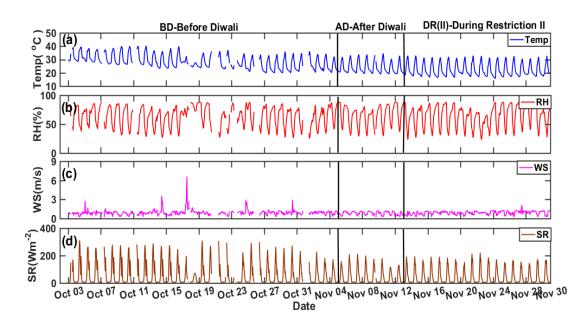


Figure 4.6: Temporal distribution of hourly average concentration of different parameters (a) Temp, (b) RH, (c) WS & (d) SR during Period II (3<sup>rd</sup> October 2021 to 30<sup>th</sup> November 2021)

Table 4.3: Statistical summary of different meteorological parameters – air temperature (AT), relative humidity (RH), wind speed (WS) and solar radiation (SR) during different study phases

Phase	Parameter	Minimum	Maximum	Mean ± S. D	Median
Before Restriction (BR)	AT (°C)	19	43	31 ± 6	29
	RH (%)	11	73	$29 \pm 13$	28
	WS (m/s)	0	6	$1\pm 0$	1
	SR (W/m <sup>2</sup> )	3	415	$79\pm104$	12
During Restriction (DR)	AT (°C)	21	44	32 ± 4	31
	RH (%)	14	88	$44 \pm 18$	42
	WS (m/s)	0	12	$1 \pm 1$	1
	SR (W/m <sup>2</sup> )	4	468	88 ± 11	17
After Restriction (AR)	AT (°C)	26	44	$34 \pm 4$	34
	RH (%)	24	88	54 ± 15	54
	WS (m/s)	0	10	$1 \pm 1$	1
	SR (W/m <sup>2</sup> )	4	406	$104 \pm 118$	29
Before Diwali (BD)	AT (°C)	20	40	28 ± 5	28
	RH (%)	25	88	63 ± 17	68
	WS (m/s)	0	7	$0\pm 0$	1
	SR (W/m <sup>2</sup> )	4	311	57 ± 77	9
After Diwali (AD)	AT (°C)	19	35	$23 \pm 4$	22
	RH (%)	29	88	64 ± 17	69
	WS (m/s)	0	1	$0\pm 0$	1
	SR (W/m <sup>2</sup> )	5	213	41 ± 53	7
During Restriction II	AT (°C)	16	33	22 ± 4	20
(DR II)	RH (%)	23	88	63 ± 18	68
	WS (m/s)	0	2	1±0	1
	SR (W/m <sup>2</sup> )	5	221	41 ± 55	7

### 4.2.5 Diurnal Behaviour of particle number concentration

The diurnal variation of the different size range pollutants shows various profiles and pattern in different seasons as well as in different hours of the day.

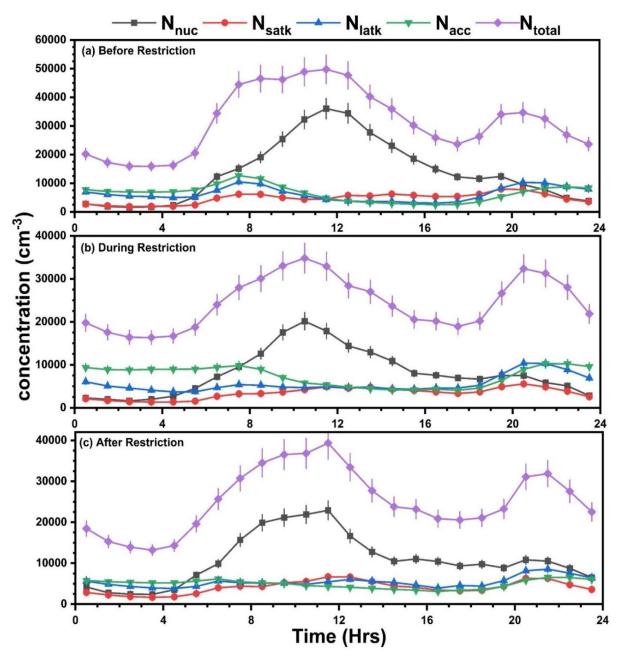


Figure 4.7: Diurnal behaviour of different size range particles during different periods, (a) Nucleation, (b) Small Aitken, (c) Large Aitken, (d) Accumulation and (e)  $N_{total}$  Concentration for different phases in period I (1<sup>st</sup> April 2021 to 30<sup>th</sup> June 2021) During the period I, diurnal concentration of the particles shows that the concentration during the vehicle rush hours namely morning and evening peak hours are not clearly

visible due to the reduced vehicular sources (Fig 4.7). The increase in concentration during the peak hours is a common phenomenon in the urban regions especially in the roadside microenvironments. The afternoon hours in the period I received a higher concentration of the particles in the lower size ( $N_{nuc}$ ) due to the natural new particle formation process (Fig 4.7). The seasonal conditions during the period I was favourable condition for the new particle formation process/UFP burst due to the optimum amount of relative humidity and solar radiation. This can be clearly seen between the morning to afternoon hours in the period I (Fig 4.7).

The diurnal variation during different periods are not uniform and it changes during different phases based on the emission and meteorology. During period II when there was no restriction for the vehicular sources, the diurnal concentration increased during peak hours. The diurnal concentration increases during peak hours 10hrs to 12hrs as well as during the evening peak hours (16 to 20 hrs) (Fig 4.8). The hours are considered as the peak hours where the vehicular flow is more compared to the normal hours of the day. The contribution of the different size ranges also varies during the different hours of the day. The N<sub>nuc</sub> concentration was found high during both BR and AR phases (37.7 % and 37.6%) which was directly influenced from the vehicular sources. In similar fashion there is another peak found during the evening peak hour due to the vehicular movement. During period II the diurnal variation throughout the period shows a double hump model pattern which is aligning with the peak hours. This shows that the vehicular sources played a major role in the atmospheric nanoparticle concentration in the study area (Kompalli et al., 2018). During the DR phase of Period I, the smaller size particles  $N_{nuc}$  and  $N_{satk}$  contributes to around 30 % and 14% in the Ntotal PNC concentration. During the same period Nlatk and Nacc contributes around 29% and 28% during the DR phase. N<sub>nuc</sub> contributes to around 30 to 40% during the different phases of period I, N<sub>satk</sub> concentration lies between 10 to 15%, N<sub>latk</sub> ranges from 25 to 30% and the  $N_{acc}$  particles concentration contributes for 20 to 30% in the period I irrespective of the phases. Similar to the diurnal concentration the percentage contribution also varies during the different periods. The percentage contribution during the period II of different size range particles are as follows: 15 to 25% for N<sub>nuc</sub> particles, 20 to 30% for  $N_{latk}$  particles, 35 to 50% for  $N_{acc}$  particles.

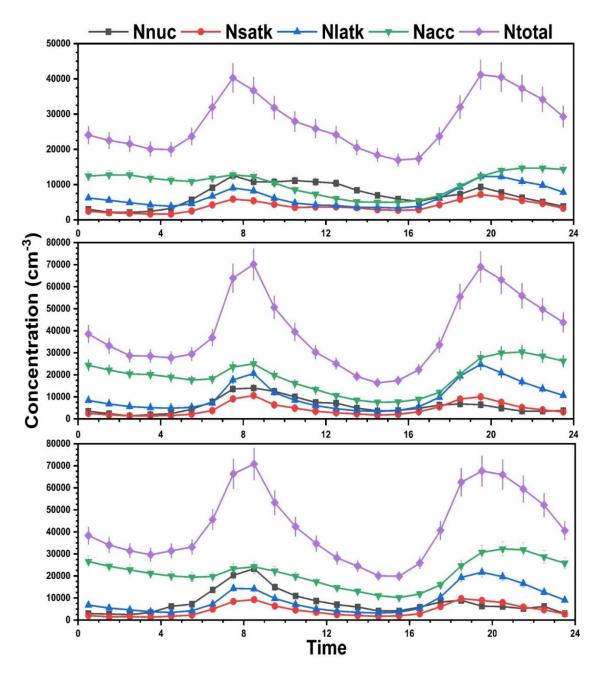


Figure 4.8: Diurnal behaviour of different size range particles during different periods, (a) Nucleation, (b) Small Aitken, (c) Large Aitken, (d) Accumulation and (e) Ntotal Concentration for different phases in period II ( $3^{rd}$  October 2021 –  $30^{th}$  November 2021)

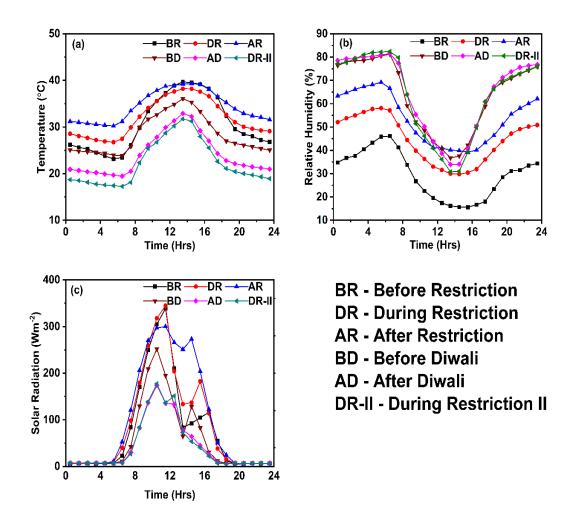
The vehicular reduction during period II was only 1% compared to the previous year and also the Diwali sources contributed for more contribution of  $N_{acc}$  sources.  $N_{acc}$  contributes to around 50% in the AD phase and 46% in DRII phase to the  $N_{total}$  concentration. The diurnal analysis reveals that the increase or decrease in

concentration in day also varies in different hours of the day. Based on the source's variation the atmospheric nano particles concentration increases or decreases. The variation is not only seen in the concentration but also seen in the size proportion. The size proportion also varies based on the nature of the emission and their contribution in the atmospheric nanoparticle's concentration.

#### 4.2.6 Impact of meteorology on particle number size concentration

The meteorological parameters of a region are another influencing parameter for determining the particle number concentration in a region. The major meteorological parameters which influences the concentration are wind speed, solar radiation and the relative humidity. These parameters help the nanoparticles to get dispersed or influenced them to undergo secondary transformations in the atmosphere (Nicolás et al., 2009; Shrestha et al., 2016; Väkevä et al., 2000b).

The relative humidity plays an important role in particle coagulation process in the atmosphere (Kumar et al., 2008b). The agglomeration of the nanoparticles causes the particles to get agglomerate from one size to another size majorly from lower to higher size under the favourable relative humidity and temperature (Cusack et al., 2013). The weather pattern in the study area varies during the different periods based on the seasonal changes of the study location (Fig 4.9). During period I the climatic condition of the study area was summer where the temperature (44.5°C) attains peak but in period II the temperature varies largely from 15.6 °C to 40.34°C (Fig 4.9). During period II the relative humidity was found high throughout the day compared to period I. The period I records maximum sunshine hours among the both periods and the maximum solar radiation was observed during the period I with an irradiance of 468 Wm<sup>-2</sup>. The diurnal analysis clearly indicates that the solar radiation and RH are inversely proportional to each other. The diurnal analysis on temperature coincides with the seasonal changes and it clearly shows that the AR phase is found to have more concentration of the temperature which occurred during the month of June which is the peak summer season in the study area. Similarly, DRII phase the period falls under the winter season which is the coldest period throughout the monitoring period. The diurnal pattern of the temperature and relative humidity are directly opposite to each other.



*Figure 4.9: Diurnal variation of (a) Temperature (Temp), (b)Relative Humidity (RH), and (c) Solar radiation (SR) of different phases during period I (1<sup>st</sup> April 2021 to 30<sup>th</sup> June 2021) and period II (3<sup>rd</sup> October 2021 to 30<sup>th</sup> November 2021)* 

## 4.2.7 Influence of wind speed on measured particle number concentration

The wind speed and direction are other major factors which helps the particles to undergo dispersion or dilution. In the urban regions that to in the road microenvironment the dispersion is mainly based on the two major factors. The first one is traffic condition, where the turbulence is created by the flow and the second one is the wind dependent where the turbulence is created by wind speed and direction. The wind direction along with the windspeed can form a vortex formation in the urban street canyon. In urban regions the wind flow can be either recirculating flow or along with the street flow (Liu and Cui, 2014). Wind parameters can contribute for the different spatial variation in particle number concentration (Hofman et al., 2016; Kozawa et al., 2012a). In the study region the majority of the wind flow is from south west to north east due to the vehicle induced turbulence. A busy road surrounds the study area in south direction. During period I the maximum wind speed recorded in the study area was 11.7 m/s and in period II it was 6.5 m/s (Fig 4.10). Bivariate polar plot analysis reveals that the pollutants are dispersed on the receptor site. The mean concentration of N<sub>total</sub> observed during the BR phase of period I found to have more concentration in the region when the wind speed was less ( $\leq 6m/s$ ). The winds travelling from the southwest to northeast is the major reason for the higher concentration from the roadside. During the DR and AR phases of period I, the wind speed in the study region was higher than the BR phase which varies from >6 m/s to < 12 m/s. During period I the wind created turbulence was more compared to the traffic induced turbulence due to the restricted traffic movement and also the recorded wind speed received was > 1.5 m/s during all the three phases. The vortex formation possibilities increase when the wind speed increases and also the exchange of particles occur from the wind induced turbulence instead of traffic induced turbulence (Kumar et al., 2008a).

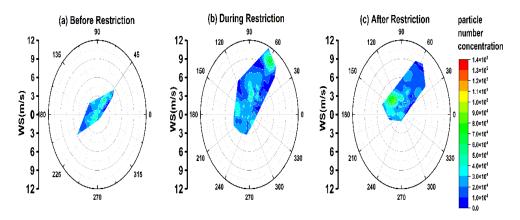


Figure 4.10: Polar plots of  $N_{total}$  with windspeed and direction during (a) BR-Before restriction, (b) DR – During restriction, (c) AR- After restriction during Period I (1<sup>st</sup> April 2021 to 30<sup>th</sup> June 2021)

Based on the velocity of the wind speed the dispersion also varied. However higher the wind speed more will be the dispersion of the particles. High dispersion of the particles resulted in less concentration (Hagler et al., 2012). Similarly, way during period II when the windspeed was less (<2 m/s) the concentration of the N<sub>total</sub> concentration of the particles were found to have higher concentration and this occurs during the AD phase. During this period the recorded windspeed recorded was from the southwest to northeast direction due to the traffic induced turbulence created in the study area. During the AD phase the vehicular sources and Diwali activities were the major emission source. The wind direction in all the different directions reveals that the PNC decreased with increase in windspeed and vice versa. The observations during the period I and period II indicates that the regions located on the leeward side of the study region was found to have more concentration due to the transport of pollutants from windward side (Kozawa et al., 2012b). This indicates that the local emissions played a significant role in PNC concentration.

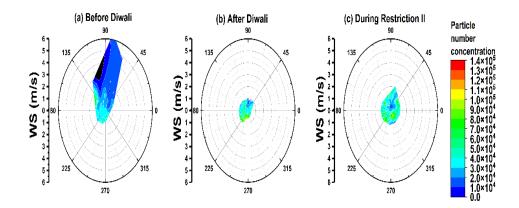


Figure 4.11: Polar plots of  $N_{total}$  with windspeed and direction during (a) BD-Before Diwali (b) AD - After Diwali (c) DR- During Restriction II during Period II (3<sup>rd</sup> October 2021 to 30<sup>th</sup> November 2021)

## 4.3 Seasonal analysis of particle number concentration

## 4.3.1 Seasonal temporal variation of particle number concentration (PNC)

The monitoring location in Delhi experienced widely varying concentration levels of particles in different seasons. The average concentration of total PNC was higher during winter, followed by spring, autumn, summer, and monsoon (Fig. 4.12). The average total PNC was found as  $>10^4$  cm<sup>-3</sup> during the year. The total PNC or N<sub>total</sub> corresponds to the sum of Nucleation - N<sub>nuc</sub>, small Aitken - N<sub>satk</sub>, large Aitken - N<sub>latk</sub>, and accumulation - N<sub>acc</sub> modes (total PNC/ N<sub>total</sub> = N<sub>nuc</sub> + N<sub>satk</sub> +N<sub>latk</sub> +N<sub>acc</sub>). The number concentration was higher during colder seasons (higher relative humidity and calm wind conditions) due to less dispersion when compared to hot seasons when wind speeds and ventilation coefficients were higher, which promoted higher dispersion resulting in comparatively lesser concentrations (Fig. 4.12, Gani et al., 2021).

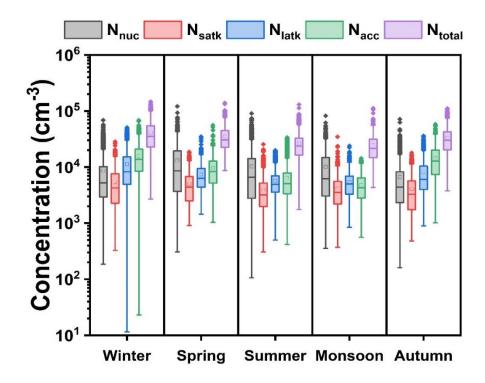


Figure 4.12: Seasonal mean concentration  $(cm^{-3})$  of particles in the nucleation  $(N_{nuc, 10} to 30 nm)$ , small Aitken  $(N_{satk}, 30 to 50 nm)$ , large Aitken  $(N_{latk}, 50 to 100 nm)$ , and accumulation  $(N_{acc}, 100 to 1000 nm)$  size range and their total  $(N_{total}, 10 to 1000 nm)$  in different seasons in Delhi

The minimum and maximum PNC in different sizes varies in the range of  $10^3$  particles cm-3 (lowest in monsoon) to >10<sup>5</sup> particles cm<sup>-3</sup> (highest in winter) (Table 4.4).

Results on a global study showed that particle concentrations may vary from  $10^2$  to  $10^7$  in highly polluted megacities such as Delhi and Beijing (Kumar et al., 2014), which are consistent with the present study. The seasonal mean PNC in different size bins of N<sub>satk</sub> (5.3 x 10<sup>3</sup> cm<sup>-3</sup>), N<sub>latk</sub> (11.2 x 10<sup>3</sup> cm<sup>-3</sup>), and N<sub>acc</sub> (15.9 x 10<sup>3</sup> cm<sup>-3</sup>) are the highest during the winter season giving rise to the highest N<sub>total</sub> in winter (exceeding 40,000 particles cm<sup>-3</sup>); interestingly, the seasonal behavior of N<sub>nuc</sub> was different than the other sizes and exhibits a peak during the spring season with a concentration of about 13.2 x  $10^3$  cm<sup>-3</sup>. The higher N<sub>nuc</sub> in spring occurs due to prevalent atmospheric conditions of moderate relative humidity and solar intensity, which facilitate the new particle formation process. The new particle formation process increases the concentration of N<sub>nuc</sub> particles significantly as only very few anthropogenic sources directly emit particles in this size range. The N<sub>total</sub> concentration in winter is ~2 times higher than monsoon and 1.6 times higher than summer (Table 3). A comparison of day and night time concentrations in different size bins reveal that the lower size range particle,  $N_{nuc}$ , is higher during the daytime (Fig. 4.13). Higher N<sub>nuc</sub> concentration in urban regions represents fresh emissions and particle formation process during the daytime, whereas higher Nacc mode particles in the nighttime occurs due to coagulation and emission from the traffic sources, especially heavy vehicles. The pattern and magnitude of Nsatk and Nlatk do not vary significantly between day and night (Fig. 4.13). The Ntotal is ~1.3 times lower in autumn and spring than winter. The lowest and highest concentrations in different sizes in seasonal scales are not uniform (Table 4.4). The PNC in bigger size range, Nacc and Nlatk, are lower during the monsoon season due to wet removal leading to lower N<sub>total</sub>, whereas PNC in smaller (N<sub>satk</sub> and N<sub>nuc</sub>) is relatively lower during the summer, monsoon and autumn seasons compared to spring due to new particle formation process and UFP bursts that occur during spring giving rise to higher  $N_{nuc}$  (Table 4.4).

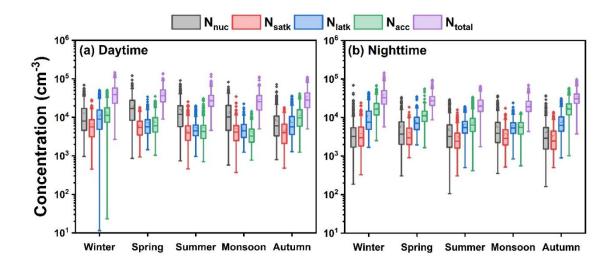


Figure 4.13: (a) Daytime and (b) nighttime variation in the number concentration of particles (cm<sup>-3</sup>) in the nucleation ( $N_{nuc}$ , 10 to 30 nm), small Aitken ( $N_{satk}$ , 30 to 50 nm), large Aitken ( $N_{latk}$ , 50 to 100 nm), accumulation ( $N_{acc}$ , 100 to 1000 nm) and total ( $N_{total}$ , 10 to 1000 nm) during different seasons in Delhi

Table 4.4 Statistical summary of particle number concentrations (PNC) classified as a function of size into nucleation ( $N_{nuc}$ ), small Aitken ( $N_{satk}$ ), large Aitken ( $N_{latk}$ ), accumulation ( $N_{acc}$ ) and total ( $N_{total}$ ) in different seasons over Delhi from the measurements made in 2020-2021

Season	Parameter	Minimum	Maximum	$Mean \pm 1\sigma$	Median
		(cm <sup>-3</sup> )	$(cm^{-3})$	(standard	$(cm^{-3})$
				deviation) (cm <sup>-3</sup> )	
Winter	N <sub>nuc</sub>	185	68626	$8594\pm9079$	5201
	N <sub>satk</sub>	327	28395	$5335\pm 3842$	4239
	N <sub>latk</sub>	11	50832	$11236\pm8719$	8248
	Nacc	23	68458	$15905\pm9645$	13803
	N <sub>total</sub>	2691	147182	$41089\pm23669$	35134
Spring	N <sub>nuc</sub>	306	121136	$13234\pm13287$	8584
	N <sub>satk</sub>	910	18674	$4981\pm 3025$	4390
	N <sub>latk</sub>	1446	35027	$7411 \pm 4385$	6326
	Nacc	1039	56033	$9599 \pm 5832$	8337

	N <sub>total</sub>	8694	139019	$35226 \pm 17909$	30413
Summer	N <sub>nuc</sub>	106	90646	$10284\pm10803$	6556
	N <sub>satk</sub>	305	25546	$3928\pm2734$	3179
	Nlatk	498	19861	$5681 \pm 3101$	4916
	Nacc	418	33838	$6411 \pm 4808$	5076
	N <sub>total</sub>	1752	130433	$26307\pm13719$	23702
Monsoon	N <sub>nuc</sub>	353	82547	$10161 \pm 10078$	6205
	Nsatk	370	34664	$4225\pm2932$	3546
	N <sub>latk</sub>	843	23892	5397 ± 2811	5010
	N acc	558	14345	$4805\pm2540$	4257
	N <sub>total</sub>	4337	112587	$24590 \pm 13256$	21459
Autumn	N <sub>nuc</sub>	161	71873	$6667\pm 6955$	4398
	N <sub>satk</sub>	479	17876	$4060\pm2917$	3263
	N <sub>latk</sub>	897	36062	$7990 \pm 5789$	6038
	Nacc	1012	58056	$14610\pm8940$	12849
	N <sub>total</sub>	3774	111316	$33329 \pm 17883$	29893

#### 4.3.2 Diurnal Variation of Particle Number Concentration

The diurnal variation of PNC in different size ranges clearly brings out the influence of traffic emissions in the urban curbside environment (Fig. 4.14). The N<sub>latk</sub> and N<sub>satk</sub> concentrations (Fig. 4.14 b, c) exhibit a more prominently clear double hump structure during the morning (07:00 to 11:00 h) and evening peak (16:00 to 20:00 h) in all the seasons confirming the dominance of traffic emissions to PNC in these size ranges in all the seasons. The N<sub>latk</sub> represents the particles in the size range of 50 to 100 nm, and most of the particles emitted from vehicular exhaust lie in this size range. In winter and autumn, N<sub>latk</sub> is higher than the PNC in the other size ranges throughout the day. Further, N<sub>latk</sub> is ~2 times higher in winter and autumn than in the other seasons. The N<sub>nuc</sub> mode particles show a similar emission pattern in all seasons with peaks (Fig. 4.14 a) during the active sunshine hours compared to the other hours of the day. This diurnal emission pattern of pollutants is induced by atmospheric and meteorological

variations in boundary layer height, temperature, wind speed, relative humidity, solar radiation and ventilation coefficient. The peak boundary layer height (BLH) (between 12:00 and 16:00 h) during winter is lowest (~500 m), whereas in summer the peak BLH increases by a factor of 4 going up to more than 2000 m in Delhi. The concentration of particles in different sizes also vary diurnally (e.g., Laakso et al., 2003).

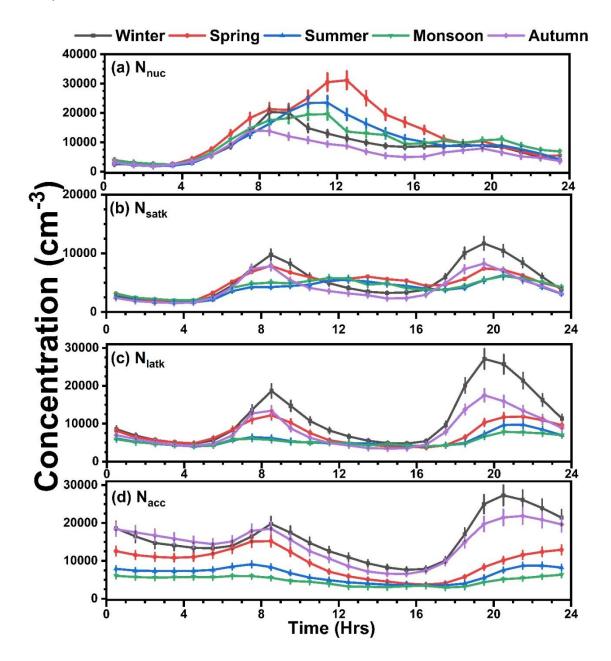


Figure 4.14: Diurnal variation (hourly average) of (a)  $N_{nuc}$ , (b) $N_{satk}$ , (c) $N_{latk}$  and (d)  $N_{acc}$  (cm<sup>-3</sup>) during winter, spring, summer, monsoon and autumn seasons in Delhi. Vertical bars correspond to  $\pm 1 \sigma$  (standard deviation) from the mean

In summer and spring N<sub>nuc</sub> is higher during the active sunshine hours owing to nucleation burst of particles, and new particle formation events, the details and the processes involved are beyond the scope of the present study. The new particle formation events start as a result of clustering particles in the sub-3 nm size range and oxidative precursor vapors. As the measurements of these events are unavailable, these events could not be probed further in this study. The SO<sub>2</sub> concentration is low during spring as it gets utilized in the formation of H<sub>2</sub>SO<sub>4</sub> vapors, which acts as a precursor for the formation of new particle in nucleation mode (Fig. 4.14 a). The concentrations of other pollutants from vehicular emissions, such as nitrogen oxides (NO, NO2 and NO<sub>x</sub>) and carbon monoxide (CO) are also higher in winter. During summer, spring, and monsoon seasons, Nnuc increases whereas Nacc decreases due to reduced condensation of particles. The microphysical process of aerosol condensation is strongly influenced by relative humidity. Over the study region, RH is lower during summer and spring than winter which leads to lower coagulation of particles. The warmer summer and monsoon seasons are usually more ventilated than the colder winters (lower BLH). The solar radiation (SR) is higher in summer whereas RH is higher in winter.

The average diurnal concentration increases during morning peak hours irrespective of the season; however, in the evening hours, the peak concentration is different in different seasons because of the differences in BLH. During the warmer periods (summer and monsoon seasons), a higher BLH result in good ventilation, enabling dilution of particles resulting in lower N<sub>total</sub>. A lower BLH, higher RH, lower temperature and lower SR helps trapping the pollutants near surface and result in significantly higher concentrations (>50,000 particles per cm<sup>3</sup>) whereas a higher BLH, lower RH, higher temperature and higher SR during summer (and in monsoon) results in at least a factor of 2-lower concentrations than winter. Thus, this synergistic analysis of PNC in different size ranges along with local emissions of gaseous precursors, processes of formation, and meteorological parameters reveals that the PNC is

governed by local emissions of gaseous precursors and meteorological parameters, and the diurnal variations in PNC exhibit seasonal changes due to variations in these parameters.

## 4.3.3 Seasonal, and temporal variation of ultrafine and accumulation range particles

The diurnal and day to day variation of ultrafine particles (UFP, particles in 10 to 100 nm size range) and accumulation mode ( $N_{acc}$ , particles in 100 to 1000 nm size range), and the total particle number concentration ( $N_{total}$  particles in 10 to 1000 nm size range) are depicted in Figure 4.15.

The peak/non peak hour changes are analyzed using box whisker plot analysis. The diurnal variation of UFP (Fig. 4.15a) during peak and non-peak hours clearly demonstrates the dominant impact of vehicular exhausts in contributing to nanoparticle concentration ( $\geq 60\%$ ) in the urban roadside microenvironment. The measurement site is located adjacent to a busy arterial road, which has a higher contribution from traffic sources, resulting in more ultrafine size range particles (<100 nm) compared to N<sub>acc</sub> (Fig. 4.15 b). Further, the day to day variation in UFP is significantly less than N<sub>acc</sub> in all the seasons confirming the higher contribution in UFP size range which is more or less constant during the year (Fig. 4.15 d, e).

The UFP concentration is highest in winter followed by spring. The peak time in UFP concentration exhibit a seasonal variation – winter, spring and autumn the peaks occur earlier than summer and monsoon when UFP concentrations are lower. Whereas  $N_{acc}$  is highest in winter followed by autumn, spring and summer with monsoon  $N_{acc}$  being the lowest. The seasonal mean concentration of UFP is the highest in spring, followed by winter, summer, monsoon, and autumn, whereas for  $N_{acc}$  mode, the order of concentration is different, it is high in winter, followed by autumn, spring, summer, and monsoon. The concentrations of particles in different size ranges vary based on the variability and intensity of sources as well as due to the local meteorological conditions of the study area (Table 4.5, Fig. 4.15).

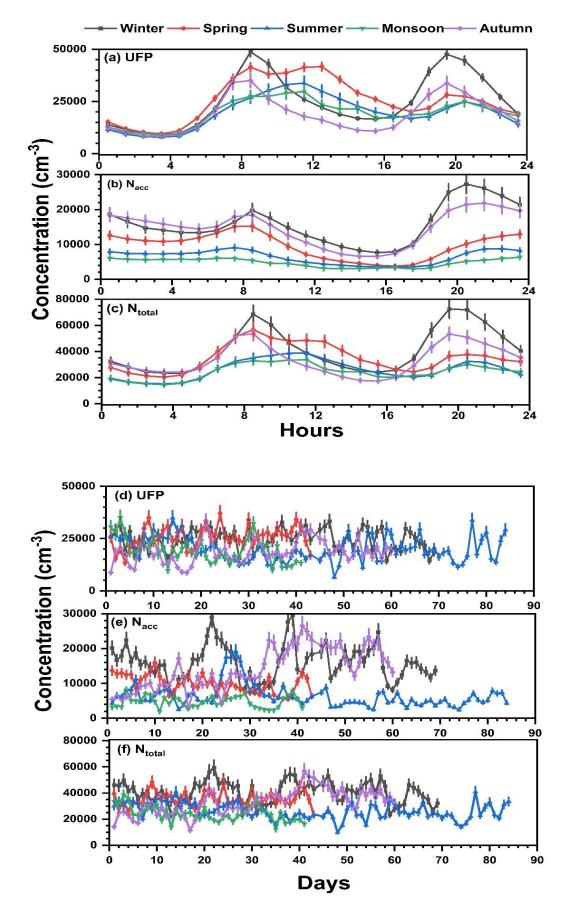


Figure 4.15: Diurnal variation of different sizes (a) UFP, (b)  $N_{acc}$  and (c)  $N_{total}$  and the temporal variation of daily mean concentration of (d) UFP, (e)  $N_{acc}$  and (f)  $N_{total}$  in different seasons during December 2020 - November 2021

Table 4.5: Statistical summary of UFP (10 to 100nm) and accumulation mode ( $N_{acc}$ ,100 to 1000nm) particles in different seasons and hours of the day over Delhi

	Hour				Accumulation mode particles (Nacc			
	of	Ultrat	fine partic	les (UFP, cm <sup>-3</sup> )	cm <sup>-3</sup> )			
Season	the							
	day*	Min	Max	Mean ± SD	Min	Max	Mean ± SD	
	MPH	19	101916	41128 ±18959	23	50785	17332 ±9171	
	EPH	12472	123505	$42869 \pm 15875$	4661	68458	26135±11303	
Winter	NPH	8	121073	$22820 \pm 14525$	3	46505	10498±6141	
	MPH	11250	123882	$39383 \pm 17588$	2601	35018	12347±5916	
	EPH	13494	677109	$26921 \pm 8924$	1806	30236	10031±4344	
Spring	NPH	4411	129904	$29836 \pm 18099$	1039	13578	5044±2157	
	MPH	6657	80490	$30037 \pm 14821$	1092	30236	6906±4494	
	EPH	3701	52382	$23210 \pm 8943$	1426	13203	7270±4446	
Summer	NPH	2281	125935	$23202 \pm 14809$	705	27585	3961±1876	
	MPH	19	70151	$27980 \pm 14324$	23	13211	4818±2646	
	EPH	8419	62950	$23354 \pm 11635$	848	11221	4988±1758	
Monsoon	NPH	5	108793	20926±15693	2	12637	3215 ±1753	
	MPH	5677	82898	27310± 16035	1376	40977	15475±7777	
	EPH	6706	64603	20947±11269	3647	55037	21023 ±1178	
Autumn	NPH	2969	48912	16182± 10121	1246	31775	8928±5437	

## 4.3.4 Peak and non-peak hour changes

The study region experiences high-ventilated periods during spring, summer and high humid conditions (Table 4.5, Fig. 4.16b) during winter, which influences the physiochemical properties of the nanoparticles (Sabaliauskas et al., 2012). The spring season is marked by frequent new particle formation events and UFP bursts, which

result in higher concentrations of UFP. However, the concentration of the UFP and Nacc particles generally show higher concentrations during the peak hours compared to the non-peak hours; during peak hours, most of the UFP concentration ranges are in 25 to  $50 \times 10^3$  cm<sup>-3</sup> (Fig. 4.16a, Table 4.4), whereas during non-peak hours the concentration decreases and it is  $<25 \times 10^3$  cm<sup>-3</sup> (Fig. 4.16b, Table 4.6) range except in spring season. The concentrations in morning and evening peak hours are not uniform throughout the study period. The UFP and Nacc during the evening peak hour is higher than the morning peak hour in winter (42 x 10<sup>3</sup> cm<sup>-3</sup> (UFP), 26 x 10<sup>3</sup> cm<sup>-3</sup> (N<sub>acc</sub>)) and in autumn (20 x 10<sup>3</sup> cm<sup>-3</sup> (UFP). 21 x 10<sup>3</sup> cm<sup>-3</sup> (N<sub>acc</sub>)) due to less ventilation/lower dispersion, higher rate of coagulation and condensation. The Nacc is the lowest in monsoon (3.2 x  $10^3$  cm<sup>-3</sup>) due to wet removal of particles in larger size. The analysis shows that the concentration of UFP and Nacc mode particles in a region is highly dependent on the sources and the local meteorological conditions of the particular region, and concentration can vary significantly under the influence of these two factors (Fig 4.17). The seasonal mean concentration of total PNC (10 to 1000 nm) in the study region varied from 24 (monsoon, the lowest) to  $41 \times 10^3$  cm<sup>-3</sup> (winter, the highest) (Table 4.6). The maximum PNC was found range of 111-147 x 10<sup>3</sup> cm<sup>-3</sup> over Delhi in different seasons; the highest PNC was observed in winter and lowest in monsoon, whereas the minimum PNC was in the range of  $1-9 \times 10^3$  cm<sup>-3</sup> with minimum value being the lowest in monsoon and highest in spring (Table 4.6).

Table 4.6: Seasonal mean concentration of total PNC covering the size range of 10 to 1000 nm over Delhi. Total PNC is given in units of  $10^3$  cm<sup>-3</sup>. Mean  $\pm 1\sigma$  (standard deviation), maximum, minimum and median values obtained in different seasons are given

Seasons	Mean $\pm 1\sigma$ (standard deviation)	Minimum	Maximum	Median
	PNC $\times 10^3$ cm <sup>-3</sup>	$PNC \times 10^3$	PNC ×	$PNC \times 10^3$
		cm <sup>-3</sup>	$10^3 \text{ cm}^{-3}$	cm <sup>-3</sup>
Winter	41.0 ± 23.7	2.6	147.2	35.1
Spring	$35.2 \pm 17.9$	8.7	139.0	30.4
Summer	$26.3 \pm 13.7$	1.7	130.4	23.7

Monsoon	$24.3 \pm 13.4$	1.4	112.5	21.0
Autumn	$33.3 \pm 17.8$	3.7	111.3	29.9

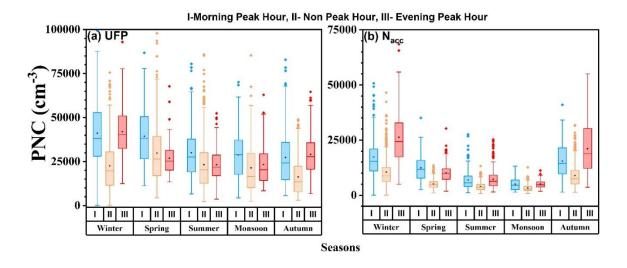


Figure 4.16: Seasonal mean concentration of particles in the (a) ultrafine range (UFP 10 to 100 nm) and (b) accumulation range ( $N_{acc}$ ) during different hours of the day – MPH - morning peak hours (7 to 11 h), NPH - non-peak hours (11 to 16 h) and EPH - evening peak hours (16 to 21 h) in different seasons during December 2020 - November 2021. The rectangular box represents data range of 25 to 75% in the box plot. The upper line represents the upper quartile value. The lower line represents a lower quartile value. The centerline of each box corresponds to the median value, and \* represents the mean value. The symbol (–I) at the top and bottom of each box denotes the maximum and minimum values

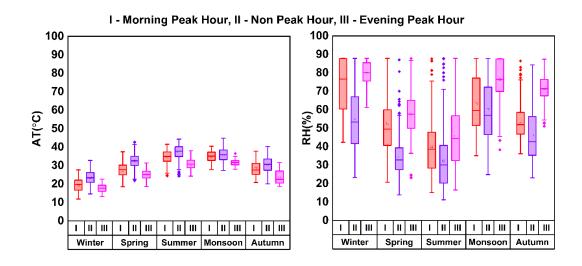


Figure 4.17: Seasonal mean concentration of meteorological parameters (a) air temperature ( $AT^{\circ}C$ ) and (b) relative humidity (RH, %) during different hours of the day. MPH - morning peak hours (7 to 11 h), NPH - non-peak hours (11 to 16 h) and EPH - evening peak hours (16 to 21 h) during December 2020-November 2021 over Delhi

### 4.3.5 Size distribution of ultrafine and accumulation mode particles

The particle number size distribution (PNSD) of measured ultrafine and accumulation mode particles are analyzed during peak and non-peak hours in different seasons to determine the dynamics of the particles (Fig. 4.18). The size distribution determines the precise particle range that dominates the season and identifies the sources similar to the concentration of particles (Fig. 4.18). An analysis of PNSD is significant as it provides quantitative information on the number of particles in different size bins, which is crucial for this study. In winter and autumn, the particle number size distribution during EPH exhibits a single clear peak concentration at around 100 nm size, which is a typical profile of vehicular exhaust sources, especially in urban regions (Sebastian et al., 2021; Yadav et al., 2021). Except for the concentration changes, the MPH follows a similar emission profile as that of EPH during winter and autumn. Besides exhaust emissions, the local meteorology prevalent during winter and autumn influences the concentration of these particles.

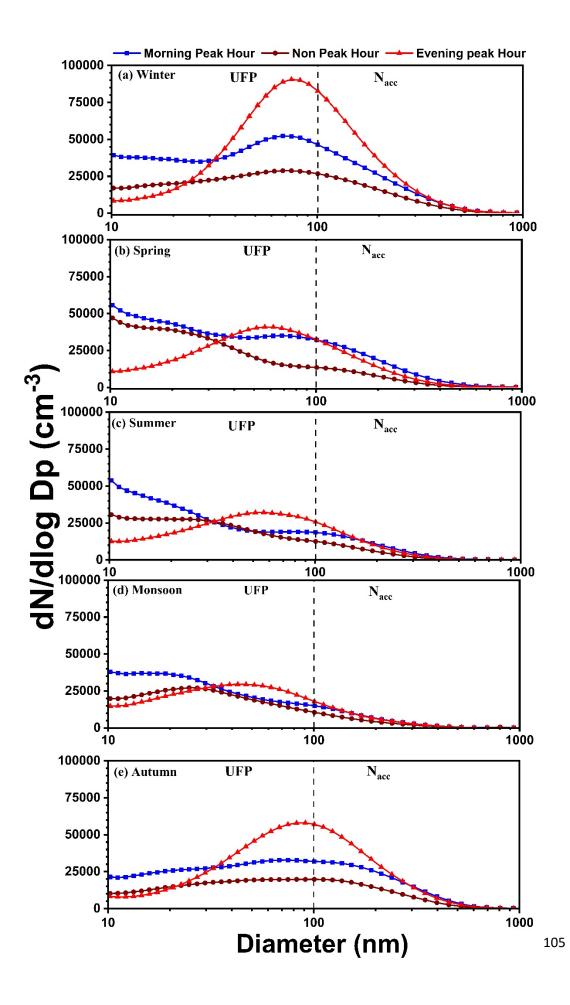


Figure 4.18: Size distribution of Ultrafine particles (UFP) (10 to 100 nm) and accumulation mode particles (100 to 1000 nm) during different hours of the day, such as morning peak hours (7 to 11 h), non-peak hours (11 to 16 h) and evening peak hours (16 to 21 h) over Delhi in (a) winter, (b) spring, (c) summer, (d) monsoon, and (e) autumn, respectively

Whereas during summer and spring (Fig. 4.18), the EPH concentration is not similar to winter and autumn due to different meteorological conditions that prevail during that period, which influences the dispersion of particles (Tyagi et al., 2020). The size distribution in the morning peak hours during summer, monsoon, and spring (Fig. 4.18 b-d) is not similar to winter. Instead, a gradual decrease in concentration from a smaller size to a bigger size occurs, leading to a higher concentration in the UFP range and a lower concentration in Nacc ranges, respectively. The increase in the concentration of smaller size range particles (< 30 nm) during spring (Fig. 5b), especially during non-peak (11-16 h), occurs due to new particle formation events driven by photochemistry, humidity, and solar intensity (which is higher, Table 3) (Deng et al., 2020; Kanawade, et al., 2021; Zimmerman et al., 2020).

## 4.3.6 Monthly and diurnal variations of particle size distribution

The monthly analysis of particle size distribution during the monitoring period reveals that the concentration of particles <300 nm is the highest in the size range of 10 to 1000 nm, and ranges between  $4 \times 10^4$  cm<sup>-3</sup> and  $5 \times 10^4$  cm<sup>-3</sup> (Fig 4.19a). During the pre-monsoon months of March, April and May the concentration of particles in the size range <50 nm is higher over Delhi, and exceeds  $4 \times 10^4$  cm<sup>-3</sup> which is attributed to gas-to-particle conversion mechanism. These months are found to be favorable for new particle formation process over the study region (Yadav et al., 2021). The annual mean total PNC over the study region is  $3.22 \times 10^4$  cm<sup>-3</sup>. The diurnal mean of PNC over the study region exhibits two peaks of higher concentration during the day – one during the morning peak hours and the other in the evening peak hours (Fig 4.19b). The pattern of morning and evening peaks remains constant throughout the year due to the dominance of particle emissions from local sources (vehicles) over the study region. The peak concentration in the evening hours is more prominent, and higher than the morning peak hour concentration due to the accumulation of particles in the atmosphere over the study region. Though the morning and evening peaks occur due to vehicular emissions predominantly, the evening peak concentration is higher owing to a significant increase in vehicular emissions accompanied with variations in meteorology and atmospheric dynamics (such as shallow boundary layer). This feature in the diurnal mean in PNC is consistent with results obtained over an urban region (e.g., Rajesh and Ramachandran, 2024).

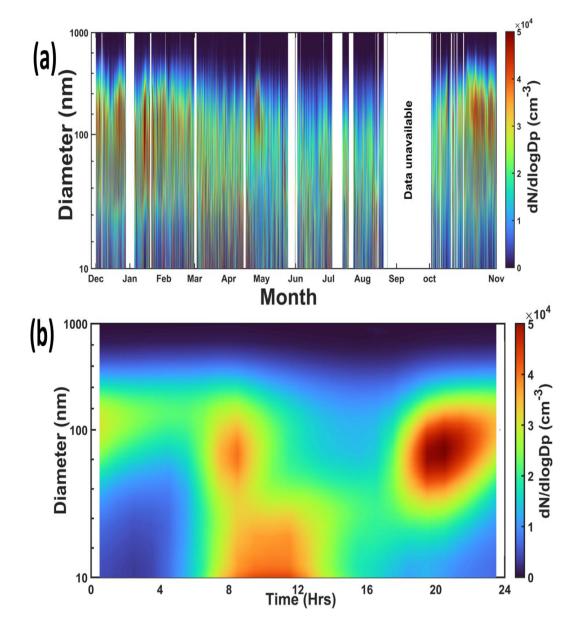


Figure 4.19: (a) Time series of particle size distribution in the size range of 10 to 1000 nm over Delhi during December 2020-November 2021. (b) Diurnal variation of particle number size distribution during the monitoring period over Delhi

#### 4.3.7 Contribution of UFP and Accumulation mode in total PNC

The contribution of UFP and  $N_{acc}$  to the total PNC varies seasonally (Fig. 4.20), due to the differences in emission sources and prevalent local meteorological conditions (mainly changes in wind speed, relative humidity, and boundary layer heights).

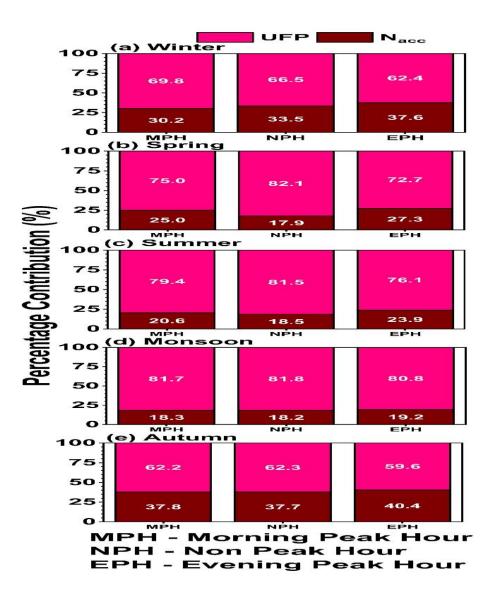


Figure 4.20: Percentage contribution of Ultrafine particles (UFP of 10 to 100 nm size) and accumulation mode particles ( $N_{acc}$  of 100 to 1000 nm size) to total concentration in Delhi during morning peak hours (7 to 11 h), non-peak hours (11 to 16 h) and evening peak hours (16 to 21 h) in (a) winter, (b) spring, (c) summer, (d) monsoon, and (e) autumn, respectively

The UFP mode (<100nm) includes particles in the nucleation (10 to 30 nm) and Aitken mode (30 to 100 nm). The majority of vehicular emissions lie in the UFP range. The meteorological factors determine the ventilation rate of nanoparticles or coagulation and condensation of the particles (Wang et al., 2017). The accumulation mode particles contribute about one-third to the total concentration in regions with higher relative humidity and calm wind conditions, such as winter and autumn (Aggarwal et al., 2012), whereas in the urban regions, the contribution of UFP to total particle number concentration is higher. In the present study, the UFP contributes about 59% to 82% to the total particle concentration in different seasons. The contribution of UFP is the highest during monsoon season (80 to 81%) (Fig. 4.20d), highest contribution observed throughout the study period, and this occurs due to the wet removal of  $N_{acc}$  mode particles to a large extent.

Similarly, during summer, the contribution of UFP is >80% (Fig. 4.20c), whereas, in summer, the reduction in  $N_{acc}$  mode particles is less as condensation and coagulation is less effective because of lesser humidity in summer. During the autumn season, UFP contributes around 59 to 62%, which is the lowest contribution observed throughout the study, and this is due to the influence of higher wind speed, which limits the aging of particles and their growth. During the study period,  $N_{acc}$  mode particles during winter and autumn contribute around 30 to 40% to the total concentration of particles, which is the highest average contribution of  $N_{acc}$  mode particles compared to other seasons. The winter and autumn seasons and the meteorological conditions are favorable for particle growth through aging and coagulation. In the study region, the sources are the same throughout the year, i.e., mostly engine exhaust sources: However, the variability in local meteorological parameters. The dominance of UFP contribution in the study region occurs due to the direct association with vehicular

exhaust. The monitoring site is located in an urban region, especially adjacent to the road, where the majority of the vehicles use petrol, diesel, and compressed natural gas as fuels, resulting in the dominance of UFP contribution to the total PNC of nanoparticles in the study region.

#### 4.3.8 Size resolved particle contribution

The N<sub>total</sub> concentration or total PNC is made up of contributions from all the size fractions (Total PNC/N<sub>total</sub> = N<sub>nuc</sub> + N<sub>satk</sub> +N<sub>latk</sub> +N<sub>acc</sub>). Apart from changes in concentration, the contribution of different size fractions to N<sub>total</sub> also varies from season to season due to changes in nanoparticle emission sources, patterns, and regional meteorology (Fig. 4.21). During the winter season, when N<sub>total</sub> was the highest, N<sub>acc</sub>, contributes the maximum at 41% followed by N<sub>latk</sub> 26% to N<sub>total</sub>. N<sub>acc</sub> contribution was higher in this season due to higher rate of coagulation in the atmosphere under high relative humidity and lower temperatures. Similarly, in autumn, N<sub>acc</sub> accounts for 45% of N<sub>total</sub>. N<sub>acc</sub> contributes  $\geq$ 30% during winter and autumn to the total. During spring, summer, and monsoon, the contribution of N<sub>nuc</sub> is 33%, 34%, and 36%, respectively to N<sub>total</sub>.

The percentage contribution of  $N_{satk}$  and  $N_{latk}$  does not varied significantly during the year in Delhi, and their contributions lie in the 12-17% range, and 22 to 26% range, respectively (Fig. 4.21). This indicates that the contribution from the emission sources of particles in the size range of 30 to 100 nm was uniform throughout the year. The smaller  $N_{nuc}$  and the larger  $N_{acc}$  exhibit a higher variation in their respective contributions to the total, varying from 20% to 36% and 23% to 45%, respectively. Overall, the contribution of  $N_{satk}$  (30 nm to 50 nm) particles to total PNC remains more or less the same during all seasons.  $N_{latk}$  particles (50 nm to 100 nm) contribute about 25% to the total PNC in all seasons (Gani et al., 2019; Patel et al., 2021).

The percentage contributions by different size particles to  $N_{total}$  during daytime and nighttime (Fig. 4.21) vary and the results clearly indicate the distinct role of diurnal variation in emission sources. The % contributions of nucleation mode followed by accumulation contribute >60% to  $N_{total}$  during daytime with their sum contribution being the highest in spring season (67%) (Fig. 4.21).

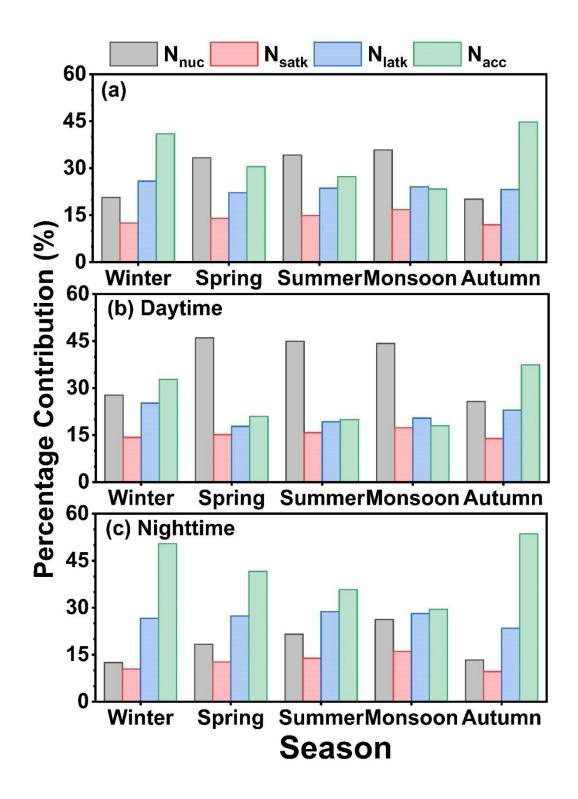


Figure 4.21: Percentage contribution (%) of  $N_{nuc}$ ,  $N_{satk}$ ,  $N_{latk}$ ,  $N_{acc}$  to total concentration ( $N_{total}$ ) from winter to autumn in Delhi (a) throughout the day (24h), (b) in daytime (08:00 to 20:00 h) and (c) in nighttime (20:00 to 08:00 h)

During the warm periods of spring, summer and monsoon,  $N_{nuc}$  contributes ~45% to  $N_{total}$  during daytime due to the prevalent meteorology conditions which enables freshly emitted exhaust emissions to remain in the atmosphere longer. Whereas in autumn and winter,  $N_{nuc}$  contributes <30% to  $N_{total}$  in daytime. In stark contrast,  $N_{acc}$  dominates the contribution (50% or more) to  $N_{total}$  during nighttime in winter and autumn as a result of coagulation of emitted particles as the prevalent atmospheric conditions (higher RH, lower temperature and BLH) favor coagulation (Fig. 4.21b). The  $N_{satk}$  contributes <20% during both day and night time in all the seasons.

The contribution from  $N_{satk}$  and  $N_{latk}$  during daytime and nighttime do not exhibit significant seasonal variabilities –  $N_{satk}$  contributes 14-17% (daytime) and 10-16% (nighttime), whereas  $N_{latk}$  contributes 18-25% (daytime) and 23-29% (nighttime), respectively. The variations in the contribution of different size particle to  $N_{total}$  suggests that the contribution from different sources either enables or inhibits due to the participation of precursor gases and prevailing meteorology, which influence the formation mechanism of nanoparticles in urban regions. Thus, it is clear that the variation in  $N_{total}$  concentration on seasonal scales occur as a result of differences in contribution from varied size ranges, sources, and meteorological conditions which vary diurnally and seasonally. Further, such complex mixture of the formation processes in different size ranges makes mitigating poor air quality conditions in these regions difficult.

### 4.3.9 Heat map analysis of seasonal particle size distribution

A heat map represents the evolution of particle number size distribution with the time in different seasons (Fig. 4.22). In winter, the intensity of accumulation mode particles is high during evening hours, especially from 18:00 to 22:00 h due to coagulation of particles and lower BLH (Fig. 4.22a). In spring, the concentration was found high between 08:00 and 16:00 hours due to gas-to-particle formation mechanism wherein the concentration of particles of size <100 nm attained a peak between 1100 and 1300 hours due to new particle formation events (Sebastian et al., 2021) (Fig. 4.22). The time periods, and favorable meteorological conditions for the new particle formation events reported in earlier studies (e.g., Gani et al., 2021) are similar as in the present

study. The scenario is similar in summer (Fig. 4.22c), however, a complete new particle formation does not occur owing to the emissions from other sources such as vehicular exhaust which suppress the new particle formation process, which was termed a UFP burst. A UFP burst refers to a disturbed or incomplete new particle formation process. During the nucleation process, accumulation mode particle concentration starts increasing gradually from the beginning of the new particle formation event until the end of the process due to the growth of particles from smaller to larger sizes after the nucleation events. This is defined as a complete new particle formation process. The new particle formation process can last for a few hours in a day. In an urban region like Delhi, emissions from different complex sources often disturb/inhibit the above formation process. During the monsoon season, the burst in smaller size particles is interrupted by precipitation events. The particle concentration changes between 08:00 and 12:00 h based on sources and prevailing meteorological conditions. The sources, and the formation of particles by nucleation and other secondary processes govern the magnitude of geometric mean diameter of particles, which was derived by analyzing the particle size distribution and the concentrations. In winter and autumn, the concentration of particles with a diameter >700 nm (that falls in accumulation mode) was higher, whereas the concentration of particles that are  $\leq 100$  nm covering the nucleation, small Aitken, and large Aitken sizes are significantly lower (Fig. 4.22). During spring and summer, the concentration of particles <700 nm was significantly higher than the smaller size particles, both of which are consistent with the formation mechanism, sources, and meteorological conditions during these seasons. In summary, the heat map provided a complete picture of formation process and concentrations of particles in different seasons on a diurnal scale, was consistent with the sources of particles and formation, and the prevailing meteorological conditions over the urban region Delhi.

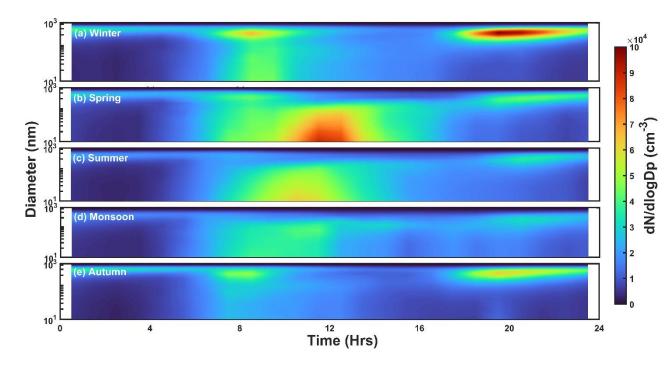


Figure 4.22: Heat map of particles in the 10 to 1000 nm size range (dN/dlogDp,  $cm^{-3}$ ) as a function of time of the day in different seasons – winter, spring, summer, monsoon, and autumn, respectively over Delhi

# 4.3.10 Number, mass, surface area and volume distribution during different seasons

The (number, mass, surface area and volume) size distributions from 10 to 1000 nm were used to determine the characteristics of measured particles (Fig. 4.23). The particle number size distribution (PNSD) showed a single clear peak during winter and autumn which was the highest measured concentration compared to other seasons (Fig. 4.23a). The peak concentration was from 100 to 200 nm in almost all seasons, which falls under accumulation mode. Generally, the particle sizes from fresh exhaust emissions of different engines were lower than 100 nm, which showed that in the study area, the emitted fresh particles frequently underwent aging processes in the atmosphere. The peak concentration in the accumulation mode also rose during humid periods due to the evaporation of aqueous droplets (Yao et al., 2007). Mass size distribution showed that the particles in the  $N_{acc}$  range were 3 to 4 times higher than other size fractions which further, contributed to the concentration of particulate matter (Şahin et al., 2022; Trechera et al., 2023). The volume of the particle size distributions

looked similar to that of the mass (Fig. 4.23b, d) since they were calculated based on the number concentration, assuming a spherical shape of the measured nanoparticles and the same density.

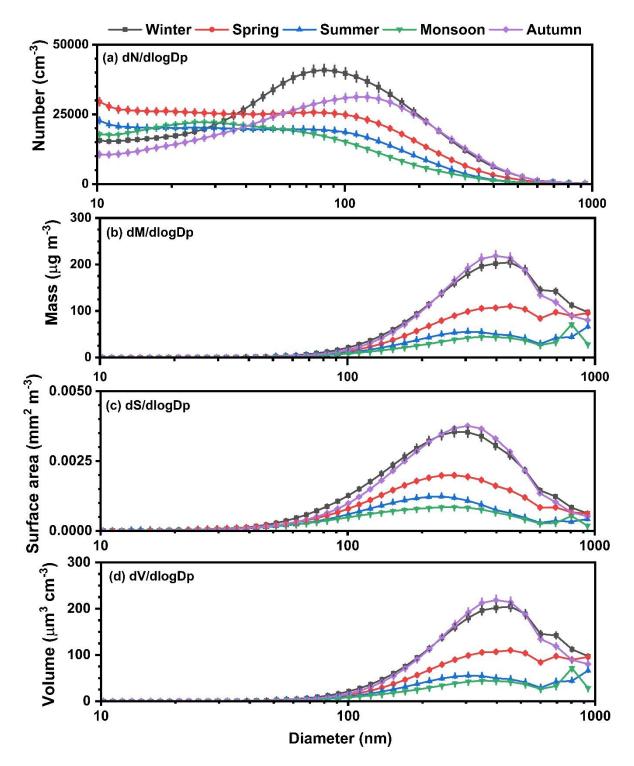


Figure 4.23: The ((a) number, (b) mass, (c) surface area, and (d) volume) size

distribution of aerosol particles from 10 to 1000 nm during winter, spring, summer, monsoon, and autumn in Delhi, a megalopolis (vertical bars represent  $\pm$  standard deviation from the mean)

Knowing the particle surface area is necessary to evaluate the health impact because the nanoparticles with lower mass and higher surface area possess a higher potential to impact health due to a higher surface reactivity (Qiao et al., 2015), and the generation of reactive oxygen species (ROS). ROS generation is an important factor in analyzing the toxic effects of the inhaled nanoparticles. It develops cytotoxicity, apoptosis, oxidative DNA damage, and cell motility (Chalupa et al., 2004; Ma et al., 2022). The analysis of the particle size distribution showed that the particle surface area was similar during the winter and autumn seasons, peaking at  $3.5 \times 10^{-3} \text{ mm}^2 \text{ m}^{-3}$ around 300 nm (Fig. 4.23c). In other seasons, even though the concentration of the particle was less, the particle surface area remained more or less the same based on the size and still posed threat to the residents.

### 4.4 Seasonal variation of gaseous pollutants

#### 4.4.1 Diurnal variation of local emissions in the roadside environment

The major source of nanoparticle emissions over the monitoring location was vehicular emissions. Apart from nanoparticles, various gaseous pollutants are also emitted due to different processes in the urban environment. The gaseous pollutants, such as CO and nitric oxide (NO), are released in the urban environment from combustion processes such as burning fossil fuels and biomass/biofuels.

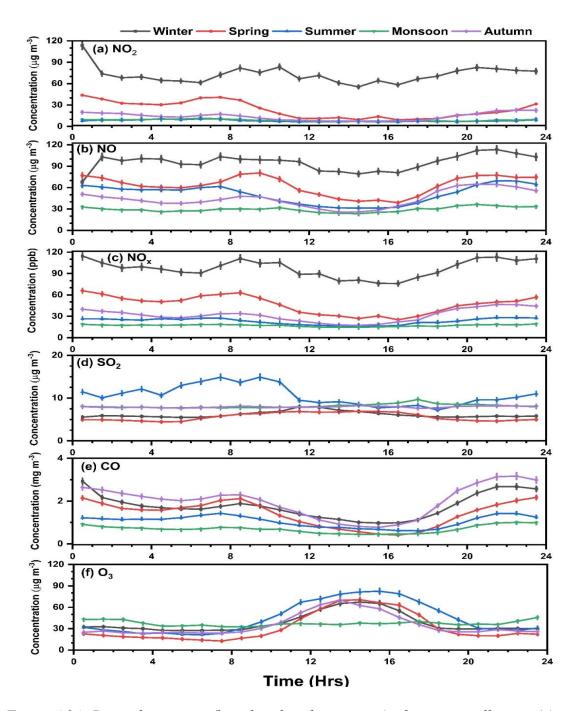


Figure 4.24: Diurnal variation (based on hourly averages) of gaseous pollutants (a) nitrogen dioxide (NO<sub>2</sub>  $\mu$ g/m<sup>3</sup>), nitric oxide or (b) nitrogen monoxide (NO  $\mu$ g/m<sup>3</sup>), (c) nitrogen oxides (NO<sub>x</sub> ppb), (d) sulfur dioxide (SO<sub>2</sub>  $\mu$ g/m<sup>3</sup>), (e) carbon monoxide (CO mg/m<sup>3</sup>), (f) ozone ( $\mu$ g/m<sup>3</sup>) during winter, spring, summer, monsoon and autumn seasons over Delhi. Vertical bars denote  $\pm 1 \sigma$  (standard deviation) from the mean

The diurnal variation of PNC and gases such as CO (and NO) shows a similar pattern, suggesting that the sources of these gaseous and nanoparticle emissions are the same, i.e., transportation sources at the monitoring location. The increase in ozone  $(O_3)$ concentration during the sunshine hours in mid-day and afternoon and reduced concentration of NOx in the presence of nitrogen dioxide (NO<sub>2</sub>) as a precursor compound due to the photochemical reaction is evident during spring, summer, and autumn when ozone formation by volatile organic compounds (VOCs) and NOx is a common phenomenon in urban regions (Fig. 4.24, Table 4.7) (e.g., Nelson et al., 2021). In winter, the concentration of SO<sub>2</sub> is relatively higher in the afternoon, suggesting that formation of H<sub>2</sub>SO<sub>4</sub> aiding new particle formation process is inhibited during this period due to unfavorable relative humidity and solar intensity (Fig. 4.24). In contrast, during summer,  $SO_2$  concentration is lower due to the use of sulfur compounds as a precursor material for the new particle formation events. During summer and spring, N<sub>nuc</sub> is also higher, suggesting that higher N<sub>nuc</sub> occurs due to the emissions and the new particle formation events that contribute to the increase in particle number concentrations in the roadside environment. The particles in the accumulation mode of size range between 100 and 1000 nm occur due to condensational growth and coagulation processes.

The particles in accumulation mode relatively have relatively longer residence times compared to nucleation mode particles. As a result, the contribution of  $N_{acc}$  is to particulate matter formation, such as  $PM_{2.5}$  and  $PM_{10}$ , is proportionately higher (Apte et al., 2011). To explore the relation between the gaseous pollutant emitted due to combustion process over an urban environment and the stable particles in the accumulation mode, carbon monoxide (CO) is correlated with  $N_{acc}$  mode (100 to 1000 nm) in all the seasons. CO is found to correlate positively well with Nacc in all the seasons in Delhi, with the coefficient of determination ( $R^2$ ) between CO and  $N_{acc}$  being  $\geq 0.48$  in all the seasons (Fig. 4.25), whereas the  $R^2$  between CO and  $N_{latk}$  (50 to 100 nm) is  $\leq 0.40$ , confirming that vehicle-based emissions are the predominant source for PNC and other gaseous pollutants in the study region.

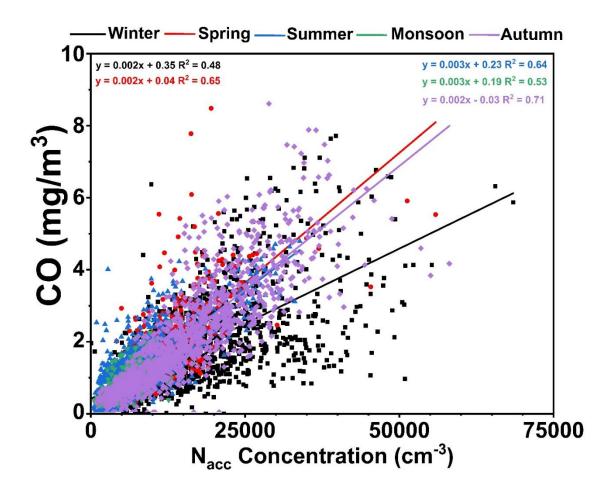


Figure 4.25: Correlation between carbon monoxide (CO, mg/m<sup>3</sup>) and particle number concentration in accumulation mode ( $N_{acc}$ , cm<sup>-3</sup>) corresponding to the five seasons over Delhi. The equations of correlation obtained for different seasons are given in respective colors in the figure

Table 4.7: Statistical summary of gaseous pollutants in different seasons over Delhi during 2020-2021

Season	Parameter	Minimum	Maximum	$Mean \pm 1\sigma$	Median
				(standard	
				deviation)	
Winter	$NO_2 (\mu g/m^3)$	5.0	488.0	$124\pm60.3$	113.4
	NO ( $\mu g/m^3$ )	0.3	489.0	$44.2 \pm 55.3$	23.5
	NO <sub>x</sub> (ppb)	0.3	456.5	$97.9\pm60.4$	81.1
	$SO_2(\mu g/m^3)$	1.8	48.0	6.2 ±3.6	4.8

	CO (mg/m <sup>3</sup> )	0.2	10.2	1.8 ±1.2	1.4
	Ozone (µg/m <sup>3</sup> )	1.1	167.8	$37.8\pm20.6$	34.7
Spring	$NO_2 (\mu g/m^3)$	0.5	297.0	$63.3 \pm 36.4$	52.2
	NO ( $\mu g/m^3$ )	5.1	291.5	$23.6\pm29.2$	12.2
	NO <sub>x</sub> (ppb)	11.4	220.0	$46.6\pm30.8$	35.9
	$SO_2(\mu g/m^3)$	2.6	41.3	$5.6 \pm 3.5$	4.6
	CO (mg/m <sup>3</sup> )	0.05	8.5	$1.4 \pm 1.1$	1.2
	Ozone (µg/m <sup>3</sup> )	0.4	172.1	31.6 ± 25.9	20.0
Summer	$NO_2 (\mu g/m^3)$	15.5	197.6	51 ± 26	44.6
	NO ( $\mu g/m^3$ )	3.9	175.4	8.1 ± 6.2	6.9
	NO <sub>x</sub> (ppb)	6.0	245.5	$23.1\pm10.8$	20.6
	$SO_2(\mu g/m^3)$	1.1	127.0	$10.6 \pm 11.3$	7.1
	CO (mg/m <sup>3</sup> )	0.1	4.8	$1.1 \pm 0.8$	0.9
	Ozone (µg/m <sup>3</sup> )	1.9	196.7	44.3 ± 33.6	34.4
Monsoon	$NO_2 (\mu g/m^3)$	10.9	211.7	$29.4 \pm 16.4$	24.8
	NO ( $\mu g/m^3$ )	2.6	36.7	$8.7\pm4.0$	7.3
	NO <sub>x</sub> (ppb)	4.2	71.9	$17.1 \pm 5.9$	15.5
	$SO_2(\mu g/m^3)$	4.6	49.0	8.1 ± 2.6	7.6
	CO (mg/m <sup>3</sup> )	0.16	2.7	$0.7\pm0.4$	0.6
	Ozone (µg/m <sup>3</sup> )	0.5	184.7	$37.4\pm30.4$	24.8
Autumn	$NO_2 (\mu g/m^3)$	7.1	319.1	44.1 ± 39.2	32.5
	NO ( $\mu g/m^3$ )	4.8	187.3	$13.9 \pm 18.2$	8.5
	NO <sub>x</sub> (ppb)	3.7	318.5	$31.7 \pm 32.5$	19.6
	$SO_2(\mu g/m^3)$	4.3	17.5	$7.9\pm1.7$	8.1
	CO (mg/m <sup>3</sup> )	0.03	11.5	$2.0 \pm 1.5$	1.6
	Ozone (µg/m <sup>3</sup> )	0.8	163.2	35.1 ± 22.6	27.6

\* NO<sub>2</sub> - nitrogen dioxide, NO - nitric oxide or nitrogen monoxide, NO<sub>x</sub> - nitrogen oxides, SO<sub>2</sub> - sulfur dioxide, CO - carbon monoxide.

The NO<sub>x</sub>, CO, and BC peak concentrations in the study region coincided with the nanoparticle's emission (Fig. 4.26, confirming that the source of these pollutants was the same (vehicular sources) (Allen et al., 2009). Studies have revealed that the transportation sector played a major role in the contribution of NO<sub>X</sub> and CO in Delhi (Bhandari et al., 2020; Pant et al., 2015; Rizwan et al., 2013). The NO<sub>X</sub> concentration was high during the winter (97 ppb) and  $\sim$  5 times higher than in the monsoon season (Fig. 4, Table 2). During the winter period, the concentration of nanoparticles and gaseous pollutants was higher in the study region. However, the concentration of N<sub>nuc</sub> was found higher in the spring season  $(1.32 \times 10^4 \text{ cm}^{-3})$  due to the gas-to-particle conversion using SO<sub>2</sub> as precursor molecule for the formation of H<sub>2</sub>SO<sub>4</sub>. H<sub>2</sub>SO<sub>4</sub> vapor acts as SO<sub>2</sub> condensation sink resulting in a lesser concentration (6 µg m<sup>-3</sup>) in the atmosphere during this season (Fig. 4.26 b). The vapor pressure of the H<sub>2</sub>SO<sub>4</sub> depends on the temperature and relative humidity. Therefore, as the spring season experienced a moderate temperature and humidity in the study region (Fig. 2.27), which favored the particle formation process. On the contrary, temperature and humidity were extremely high or low during the summer and monsoon seasons (Fig. 4.27), respectively, inhibiting the SO<sub>2</sub> condensation sink formation. Co-pollutants (O<sub>3</sub>, BC) measured in the study region were used to identify the sources and the influence of the atmospheric dynamics on the nanoparticle concentration. The O<sub>3</sub> concentration showed a mid-day peak in almost all the seasons (Fig. 4.26 d) due to the photochemical formation at ground-level from primary pollutants (Nelson et al., 2021; Sharma et al., 2013). BC was used as a tracer for local traffic emissions in the study area, exhibiting higher concentrations during the night than daytime (Fig. 4.26 e) which was consistent with the diurnal variation of Nacc.

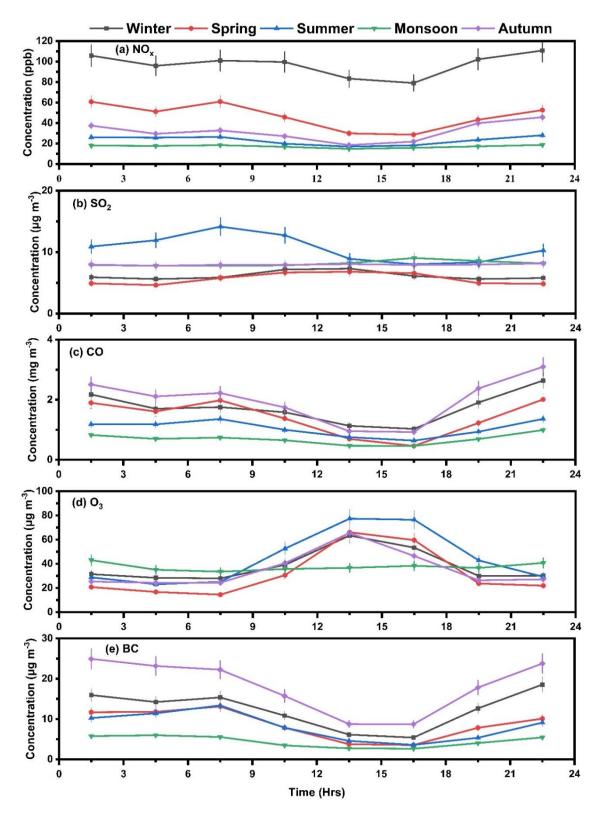


Figure 4.26: Diurnal variation of (a)  $NO_X$ , (b)  $SO_2$ , (c) CO, (d)  $O_3$ , and (e) BC during winter spring, summer, monsoon and autumn seasons in Delhi (vertical bars represent  $\pm$  standard deviation from the mean)

### 4.5 Seasonal variation of meteorological parameters and their role in PNC

#### **4.5.1 Diurnal variation of the meteorological parameters**

The concentration of particles in urban regions exhibit seasonal variability during the day due to the differences in meteorological parameters and emission sources (Sengupta et al., 2022). The particle concentration become higher during winter when the temperature was minimum and relative humidity was maximum (Fig. 4.27). The seasonal mean temperature during winter was 19°C during which the total particle number concentration was the highest. The lowest concentration of particles occurs in the monsoon season due to wet removal of particles by precipitation. Solar radiation and relative humidity are other important factors that aid particle formation and secondary pollutant formation. During the afternoon hours in almost all the seasons, the solar radiation increases, and the relative humidity decreases (Fig. 4.27). During spring, summer and autumn in the afternoon hours, the increase in solar radiation increases the secondary ozone formation and decreases the particle number concentration due to dispersion. The average solar radiation during the peak sunshine hours was around 870 Wm<sup>-2</sup> during summer (Fig. 4.27) and lowest in winter (535 Wm<sup>-</sup> <sup>2</sup>) (Fig. 4.27). Relative humidity was <50% during summer and in winter it was highest 95%. During spring, monsoon and autumn, the temperature and relative humidity were higher, and found in the range of 25°C to 32°C, and 54 to 71%, respectively. In general, the particle concentration peaks when the daily mean temperature was minimum and slumps when the daily mean temperature was high. This was due to the influence of the atmospheric boundary layer that varies with surface temperature; hence, the dispersion also varied accordingly (Bhandari et al., 2020; Gani et al., 2021). In winter, the particle concentration usually peaks due to temperature inversion; in summer, it was lower due to the strong air exchange. During nighttime, the surface temperature was lower, and the atmospheric BLH is also low, leading to a higher accumulation of particles. The meteorology plays a significant role in determining the nanoparticle concentration in the atmosphere. The planetary boundary layer height (PBLH) and ventilation coefficient directly influenced pollution concentration at ground level (Kompalli et al., 2018; Sorribas et al., 2015). During the winter, when the PBLH was observed lower than 1000 m (the lowest of all seasons) (Fig. 5c) and wind speed (Fig.

4.27f) was  $\sim 1 \pm 0.5$  m/s, resulting in lower ventilation coefficient (600 m<sup>2</sup>s<sup>-1</sup>, Table 4.8) (Fig. 4.27d), the highest mean average concentration of nanoparticles was observed (4.11 x 10<sup>4</sup> cm<sup>3</sup>). During summer and spring, the wind speed and PBLH were high, and the resultant ventilation coefficient also reached the maximum (3500 m<sup>2</sup>s<sup>-1</sup>).

Season	Parameter	Minimum	Maximum	Mean $\pm 1\sigma$	Median
				(standard	
				deviation)	
Winter	T (°C)	10.7	33.7	$19.0 \pm 4.5$	18.4
	RH (%)	25.0	88.0	$72.8\pm17.5$	80.6
	WS (ms <sup>-1</sup> )	0.1	9.0	$1.0\pm0.5$	1.0
	SR (Wm <sup>-2</sup> )	4.3	278.4	$40.5\pm55.3$	8.0
Spring	T (°C)	16.9	42.7	$26.6\pm5.5$	25.8
	RH (%)	13.8	87.9	$54.5\pm20$	54.1
	WS (ms <sup>-1</sup> )	0.1	10.0	$1.1\pm0.9$	1.0
	SR (Wm <sup>-2</sup> )	4.3	383.6	$62.8\pm83.6$	7.8
Summer	T (°C)	18.5	44.3	$32.5\pm5.2$	32.0
	RH (%)	11.2	87.8	$44.1 \pm 18.1$	42.1
	WS (ms <sup>-1</sup> )	0.2	11.7	$1.4 \pm 1.3$	1.0
	SR (Wm <sup>-2</sup> )	3.4	468.1	$91.4 \pm 114.2$	19.7
Monsoon	Ozone ( $\mu g/m^3$ )	0.5	184.7	$37.4\pm30.4$	24.8
	T (°C)	27.4	44.8	$33.0\pm3.4$	31.9
	RH (%)	24.9	87.8	$71 \pm 16.4$	74.4
	WS (ms <sup>-1</sup> )	0.2	11.0	$1\pm0.8$	0.9
	SR (Wm <sup>-2</sup> )	5.5	393.2	$73.1\pm96.9$	18.0
Autumn	T (°C)	15.6	40.3	$25.9\pm5.7$	25.5
	RH (%)	23.1	87.9	$63.5\pm17.9$	67.9
	WS (ms <sup>-1</sup> )	0.2	6.6	$0.9\pm0.4$	0.9
	$SR(Wm^{-2})$	4.4	311.5	$49.8\pm68.1$	7.4

Table 4.8: Statistical summary of the meteorological parameters

T - Temperature, RH - relative humidity, WS - wind speed and SR - solar radiation.

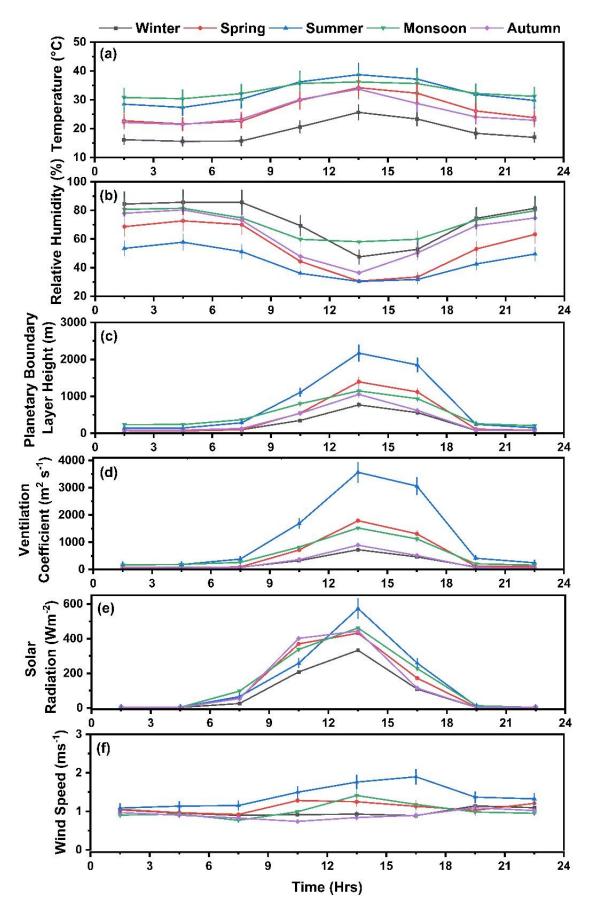


Figure 4.27: Diurnal variation of (a) ambient temperature, (b) relative humidity, (c) planetary boundary layer height, (d) ventilation coefficient, (e) solar radiation, and (f) wind speed during winter, spring, summer, monsoon and autumn seasons in Delhi

#### 4.5.2 Influence of relative humidity and temperature on particle size distribution:

The analysis revealed that temperature and relative humidity influenced the size distribution of particles largely through the processes of condensation and coagulation (Figs. 4.28, 4.29). The concentration of particles in the lower size range, especially the nucleation mode, was significantly higher during low RH (<30%) (Fig. 4.28a) and high temperature (T>30°C) (Fig. 4.29c) conditions that prevail during spring in the study area which favors the particles to undergo the natural gas-to-particle conversion mechanism. The mid-temperature range (20°C- 30°C) and the mid-RH (30% - 60%) acted as a transition period during which the lower size particles undergo physical transformation in the atmosphere and their size distribution gradually changes increasing from smaller to the higher size. In the above RH and temperature regimes, the particle concentration ranged between 1.5 and  $3.0 \times 10^4$  cm<sup>-3</sup> in all the three modes (N<sub>nuc</sub>, N<sub>satk</sub>, and N<sub>latk</sub>) irrespective of the season. In the lower temperature regime (<20°C) and higher RH (>60%) scenario, the concentrations of particles were higher in the accumulation mode size range (Fig. 4.28c, 4.29a). During the winter period, when the relative humidity was higher than 60% and the temperature was lower than 20°C, the concentration of accumulation mode particles was also found higher due to the rapid condensation and coagulation of the engine exhausts (Dinoi et al., 2023). The natural gas to-particle formation process that occured during the warmer period led to a substantial increase in nucleation size particles. During this period, relative humidity remained low compared to the winter one.

Under higher RH, condensation process begins which increases the number concentrations in the accumulation mode, and this phenomenon is evident in the winter and autumn seasons, followed by the spring season. The analysis clearly indicates that apart from the intensity of emission sources over the study region, the particle number concentration and size distribution exhibit significant variations based on the prevalent meteorology, especially relative humidity and temperature.

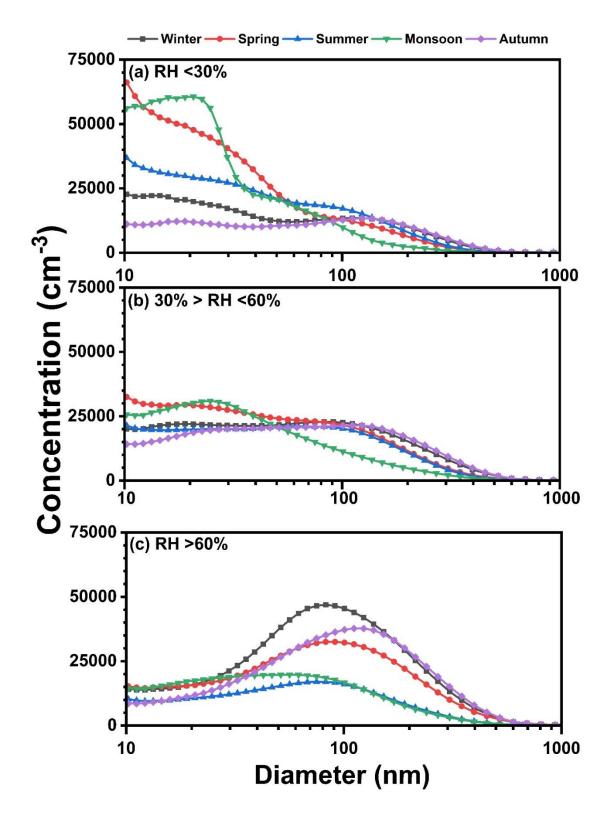


Figure 4.28: Particle number size distribution characterized for different conditions of relative humidity (RH) as (a) RH<30%, (b) 30%>RH<60%, and (c) RH>60% over Delhi in different seasons of the study period

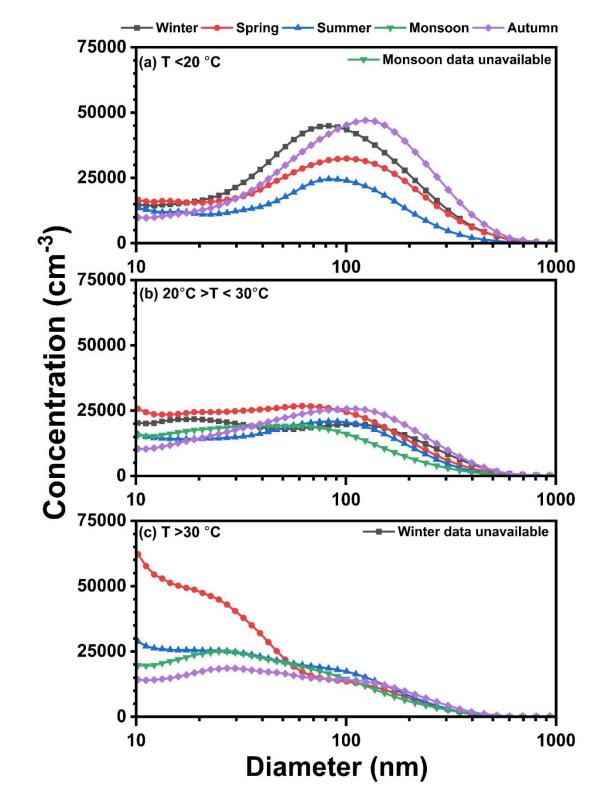
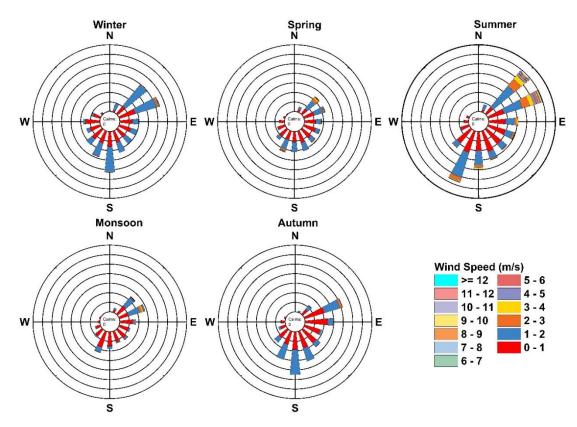


Figure 4.29: Particle number size distribution classified as a function of temperature (T, °C) over Delhi as (a) T < 20°C, (b) 20°c > T < 30°C, and (c) T > 30°C in different seasons of the study period

## 4.5.3 Role of wind speed and direction on PNC

Wind and traffic induced turbulence determined the nanoparticle concentration in the study region (Mehel and Murzyn, 2015), in the same way as the direction of wind flow. The road where the monitoring location was located, is an urban street canyon, thus, wind recirculation is common. In summer, when wind speed was higher than 2 ms<sup>-1</sup> (Fig. 4.30) during the mid-day time, N<sub>nuc</sub> and N<sub>acc</sub> particle concentrations were 2.0 x  $10^4$  and 0.75 x  $10^4$  cm<sup>-3</sup> respectively, being the lowest concentration measured among all seasons. This revealed that higher wind speeds increase the dispersion rate of particles. The dispersion rate may vary based on the particle size and the wind intensity.



*Figure 4.30: Wind rose diagram for the study area during the different seasons (winter, spring, summer, monsoon, and autumn)* 

The wind rose diagram showed that most of the wind flow in the monitoring site is from South-West towards North-East direction (Fig. 4.30). The monitoring location is covered with the arterial road on the south, and the monitoring station was

located north of the arterial road. The monitoring location was located on the leeward side of the arterial road which caused the dominance of windflow in the same direction throughout the year due to the country's driving pattern. The left-hand driving pattern of vehicles in W-E direction caused a traffic induced wind turbulence of S-W to N-E direction (Fig. 4.30) which is clearly visible throughout the study period.

## 4.5.4 Role of precipitation on PNC and PNSD

Apart from relative humidity, temperature, and wind parameters, precipitation plays a major role in determining the concentration of nanoparticles and their size distribution in the atmosphere. Thus, here, the particle size distribution was analysed before and after the precipitation to know its impact on number and size.

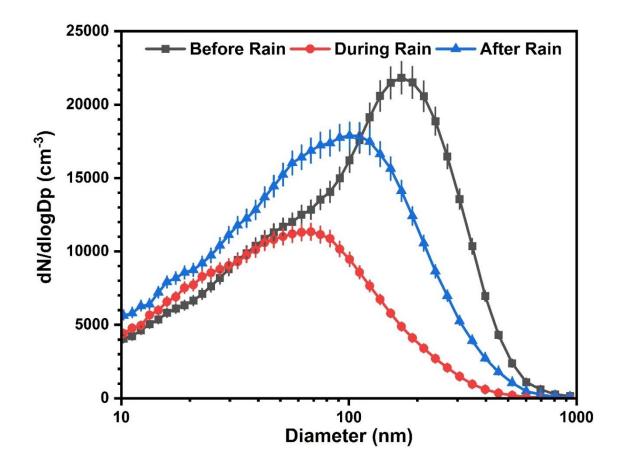


Figure 4.31: Particle number size distribution from 10 to 1000 nm during different phases of precipitation: before (17 October 2021), during (18 October 2021), and after

precipitation (19 October 2021) in the monitoring station in Delhi (vertical bars represent  $\pm$  standard deviation from the mean)

Time-resolved particle concentration with respect to the nanoparticle size revealed that precipitation-induced wet scavenging was directly associated with particle removal, and that during the precipitation period the particle concentration was reduced to half of pre-precipitation one  $(2.2 \times 10^4 \text{ to } 1.1 \times 10^4 \text{ cm}^{-3})$  (Fig. 4.31). The precipitation removed larger particles, especially N<sub>acc</sub> (100 to 1000 nm), from the atmosphere (Andronache, 2004). Thus, the particle number size distribution during the post-precipitation showed a lower concentration in this range of particle sizes. During a normal day, when there was no precipitation, the hourly average concentration of the particles ranged from 5 to 7 x  $10^4$  cm<sup>-3</sup> during the early morning hours (1 to 5 AM), especially in larger particle sizes (Nacc) (Fig. 4.32 a). This high concentration was the result of the coagulation processes among atmospheric particles (Fig. 4.32a). On the study day (18 October 2021) a precipitation of 1 to 3 mm occurred during the early morning hours when the concentration was observed maximum in the monitoring station. This resulted in a reduction of particle concentration, especially in the highestsize (N<sub>acc</sub>), reaching ~ 3 x  $10^4$  cm<sup>-3</sup> (Fig. 4.32 b). The precipitation process reduced the total concentration of the nanoparticles in the atmosphere (Fig. 4.33).

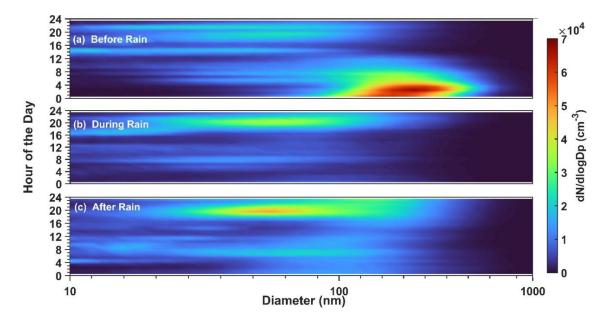


Figure 4.32: Time-resolved particle number size distribution from 10 to 1000 nm on consecutive days: day before (a) (17 October 2021), (b) day with (18 October 2021), and (c) day after (19 October 2021) precipitation in the monitoring station in Delhi

The concentration of  $N_{nuc}$  particles was found higher than larger-sized particles due to its greater diffusivity (Fig. 4.33a). The  $N_{atk}$  particles suffered greater diffusion during less-intensity rainfall, decreasing its concentration when the intensity of rainfall increased (Fig. 4.33b). In general, the particles in  $N_{atk}$  and  $N_{acc}$  were more influenced by the precipitation processes (Fig. 4.33b, c). Post-precipitation analysis showed a higher growth rate of particles, even though the concentration was less. In this regard, the particle growth was induced by the moisture content, which causes rapid condensation of smaller particles (Zhao et al., 2021).

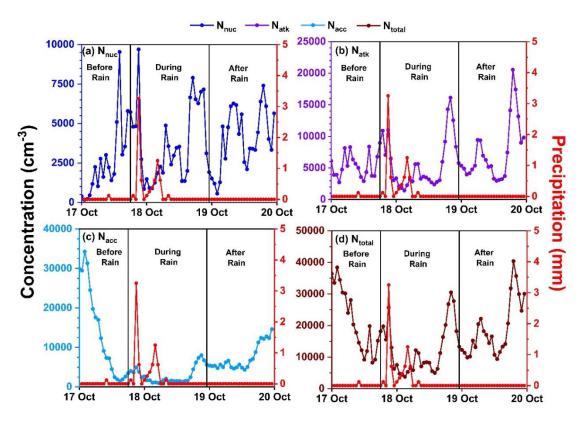


Figure 4.33: Temporal variation of particle number concentration before, during and after precipitation period for (a)  $N_{nuc}$  - 10 to 30 nm, (b)  $N_{atk}$  - 30 to 100 nm, (c)  $N_{acc}$  - 100 to 1000 nm and (d)  $N_{total}$  - 10 to 1000 nm from 17<sup>th</sup> October to 20<sup>th</sup> October 2021, in the monitoring station in Delhi

#### 4.6 Health impacts assessment

Epidemiological studies have examined the relationship between exposure to air pollution and various human health issues (Sharma et al., 2023). Air pollution was the fourth leading risk factor for early death worldwide in 2019, according to the *State* of Global Air (Health Effects Institute, 2020). The particulate matter PM2.5 was ranked as the fifth mortality risk factor in 2015, and around 4.2 million premature deaths were recorded in 2016 in low and middle-income countries globally (Manojkumar and Srimuruganandam, 2021a). Various control measures have been taken in the last few decades, such as emission norms, Bharat stage regulation (BS VI), odd-even vehicle policy schemes (Mishra et al., 2019; Tiwari et al., 2018), Graded Response Action Plan (GRAP) (Singh and Kulshrestha, 2020), and National Clean Air Action Plan (NCAP) (Sulania and Singh, 2019), have worked positively towards a decline in the concentration of particulate matter in India. However, on the other hand, UFP concentrations are increasing gradually (Sebastian, et al., 2021; Sharma et al., 2023) and emphasizing the need for further studies on UFPs. UFP also majorly affects morbidity, leading to premature death (Donaldson et al., 1998). The particles deposited in the respiratory system can interact with cells and bio-molecules. Therefore, evaluating the deposition rate of these particles in the respiratory system is crucial. The deposition rate of particles in different regions of the human respiratory system can be determined with size-segregated measurements of particles using a Multiple-Path Particle Dosimetry (MPPD) Model (Fig 4.34). This model is being widely used successfully for stimulating the deposition of size-segregated particles in the respiratory system. The unavailability of size-segregated data measurements is a significant limitation in this domain globally; however, recent advancements in instrumentation have enabled the measurement and analysis of size-segregated data. When the concentration of smaller-size particles is higher in the atmosphere, the deposition was found to be in the order of alveolar > bronchiolar > trachea bronchiolar > head regions (Ma et al., 2022). Few studies have analyzed the distribution of nanoparticles, their behavior, dynamics, and deposition potential in urban regions in different microenvironments. Estimation of particles during different vehicle fleet

hours along with the deposition potential for the commuters in the road microenvironment remains one of the potential and important areas to be explored.

#### 4.6.1 Inhalable nanoparticle concentration

The Inhalable particle number (IPN) concentration is a mathematical estimation of an individual's exposure to a concentration of particles during various physical activities that have different inhalation volumes (Prabhu et al., 2019; Qiu et al., 2019). This method is considered as one of the easiest and reliable methods for analyzing the concentration of inhalable particle concentration for different rates of inhalation. The method uses the number concentration of the particles rather than the mass concentration, and as a result, captures better the health effects as it accounts for the number of particles that penetrate deeper into the alveoli and bronchioles (Kim et al., 2017; Koehler and Peters, 2015). The equation used for this calculation is given in equation below.

$$IPN(min^{-1}) = PNC(cm^{-3}) \times IR(cm^{3}min^{-1}) - (15)$$

In equation, IPN represents the inhalable particle number, PNC represents the particle number concentration of nanoparticles, and IR represents the inhalation rate. Different IR rates are based on US EPA exposure guideline handbook, and IR rate varies based on different physical activities - light activity like walking results in less IR, and heavy activities such as running and weight lifting result in a higher IR. The deposition of air pollutants in the lungs during heavy exercise is 4.5 times higher than in the normal sitting or resting position (Ma et al., 2022). The advantage of this method is that it is the simplest mathematical method for calculating the inhalable particle number, and only the concentration of the particles is sufficient for calculation. The limitation of this method is that it provides only the concentration of the inhalable particle number with no significant insights into the actual concentration of the particles deposited in the respiratory system.

#### 4.6.2. Particle dosimetry model

The deposition of ultrafine particles and other size range nanoparticles in the human respiratory tract (HRT) can be evaluated using an MPPD model version 3.04.

The Hammer Institute of Health Science, USA designed and developed this model. The dosimetry model is a well-validated software developed by Applied Research Associates, which is considered one of the most accurate models for dosimetry analysis and works based on computational fluid dynamics. More details about the model can be found at https://www.ara.com/mppd/. Further, this model is well-validated, tested, and widely used for research and education to estimate particle deposition in the human respiratory tract (Khan et al., 2022; Manojkumar and Srimuruganandam, 2021b, 2022b). It is considered as one of the best available models for analyzing deposition in the respiratory tract. The advantage of the model is that it can be used for a wide size range of particles starting from 0.001  $\mu$ m (1 nm) to up to 100  $\mu$ m (1x10<sup>5</sup> nm). The model can be used for different species, such as rats, rabbits, pigs, and humans. The major input parameters of the model are breathing frequency (BF), tidal volume (TV), upper respiratory tract volume (URT), and other parameters such as density, size distribution, and exposure conditions.

In the MPPD model, the exposure assessment can be customized for different exposure conditions under different assumptions, such as particles entering the respiratory tract via the nose through upright body orientation and dispersion. The particles are also assumed to be in multi-particle distribution mode. According to the model recommendation, the other model assumptions and constants are used from similar global studies. After incorporating the necessary assumptions, the model is used for simulation, and the results obtained on the concentration of nanoparticles deposited in different regions of the respiratory tract, such as the pulmonary, tracheobronchial, and alveolar regions, are analyzed. The simulations are done for different seasons, and peak and non-peak hours when conditions are quite different. The advantage of this model is that it can predict particle deposition in different regions of the respiratory tract, such as the alveoli, bronchus, and bronchiole. The model uses both theoretical and numerical fittings, an added advantage. The limitation of the model is that it requires size-segregated measurement data for running simulations, and another limitation is that predicted values of regional deposition might differ from actual deposition inside the respiratory tract. The variation occurs due to water vapor, which condenses on the inhaled particles and influences the growth of particles in the biological tract of the human being inside the lungs under humid conditions.

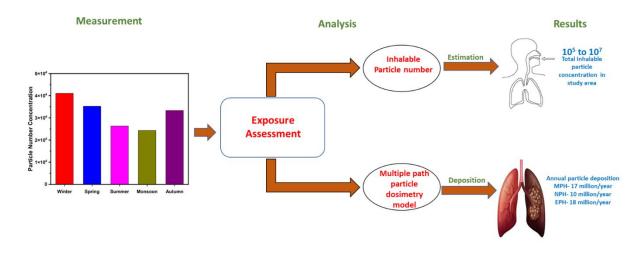


Figure 4.34: Pictorial representation of the health impact analysis methodology

# 4.6.3 Estimation of inhalable particle number concentration

The inhalable particle number (IPN) of different size fractions (UFP and total) calculated using equation (15) are plotted in Fig. 4.35. The total inhalable particle numbers determine the maximum possible exposure to different sizes of nanoparticles in the urban roadside environment, especially for motorists, drivers, police personnel, street vendors, and other people residing/working in the vicinity of the road. The exposure concentration of IPN in the study area varies from 1 million  $(10^6)$  to 10 million  $(10^7)$  particles per hour. The concentration varies based on the concentration of nanoparticles and inhalation rate (Fig. 4.35). In the spring season, the UFP shows a maximum inhalation of around  $10^6$  particles/hour under the severe inhalation category, which consumes about 100 cm<sup>-3</sup>/min air volume, followed by winter (Fig. 4.35). The exposure rate is high in spring for all kinds of exposure activities followed by winter (Fig. 4.35). The inhalable particle concentration is directly proportional to the inhalation rate and the mean concentration of particles during a particular season. In semi-urban cities such as Dehradun in India, the inhalable particle concentration was around 200 million from the fireworks during the festival (Deepavali) (Prabhu et al., 2019). The increase in inhalable particle numbers during different seasons increases the vulnerability of humans. Ultrafine and nanoparticle exposure is higher in patients

with asthmatic disorders than in normal people. The prolonged exposure of these particles causes more deposition of particles in the pulmonary and alveolar regions. Exposure to particles at higher concentrations can easily reach the lungs. From there, the inhaled particles can penetrate to other organs through lung vasculature or mobile cells (Schraufnagel, 2020).

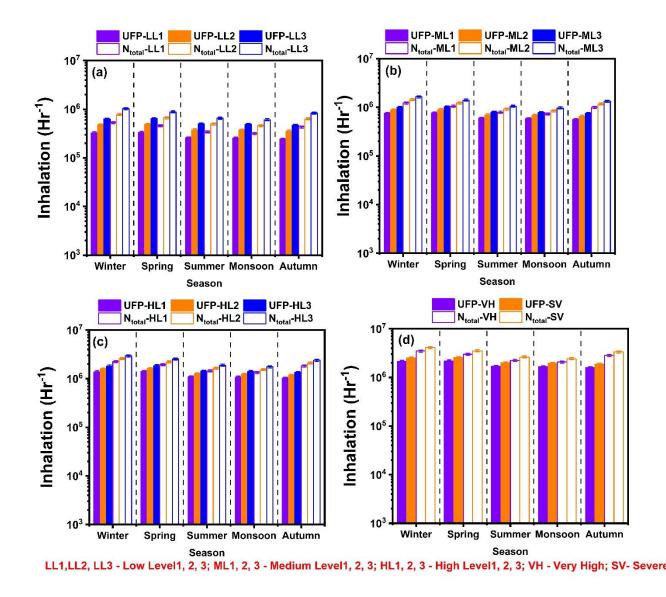


Figure 4.35: Inhalable particle number concentration for UFP (10 to 100 nm) and total particle number concentration ranging from 10 to 1000 nm (Total PNC) during different seasons for different inhalation volumes, a) light physical activity, b) medium physical activity, c) heavy physical activity, and d) very heavy and severe physical activity.

During higher IPN concentrations, populations at risk, such as the elderly and children, activity on the roadside can be restricted. The role of particulate matter is well known for the mortality and exacerbations in persons with chronic obstructive pulmonary disease (COPD), and exposure to number concentration of particles having lower mass and higher surface area is associated with stroke, ischemic heart disease, myocardial infarction, thrombotic stroke, and hypertension (Schraufnagel, 2020). The correlation between total mortality and cardiovascular mortality is high when the particle size decreases, showing that exposure to UFP/N<sub>acc</sub> and overall total concentration are highly associated with various health impacts (Schraufnagel, 2020). The results from the study corroborate the above findings and emphasizes that these quantitative results will be crucial to devise measures to help improve public health.

# 4.6.4 Quantification of nanoparticle deposition in human lungs during peak non peak hours

The seasonal mean concentration of total PNC in the size range of 10 nm to 1000 nm during different seasons and different hours of the day is utilized to predict the deposition rates in different lung regions using the MPPD model (Fig. 4.36). The particle concentrations are classified into MPH, EPH, and NPH based on the vehicular fleet in the study region. The deposition of these particles in different regions and overall lung deposition is simulated using model parameters such as Functional Residual Capacity (FRC) = 3300 ml, Upper Respiratory Tract (URT) volume = 50ml, and the density of particle =  $1.63 \text{ g/cm}^{-3}$  (Gani et al., 2020), constant exposure condition, with upright body orientation and nasal breathing scenario. The constant values in the model are kept constant for all the seasons, and the concentration of nanoparticles at different hours of the day have been used to analyze the deposition as per the methodology adopted in previous studies (Manoj Kumar et al., 2019; Manoj Kumar and Srimuruganandam, 2022a). Constant values of FRC (3300 ml) and URT (50 ml) are utilized following the model recommendation and the various studies performed on the human respiratory system. Based on the variation in the concentration of particles, the exposure and deposition vary season to season (Fig. 4.36). The output of the model represents the deposition of particles in human lungs

similar to that of computerized tomography (CT) scan results of lung deposition (Khan et al., 2022).

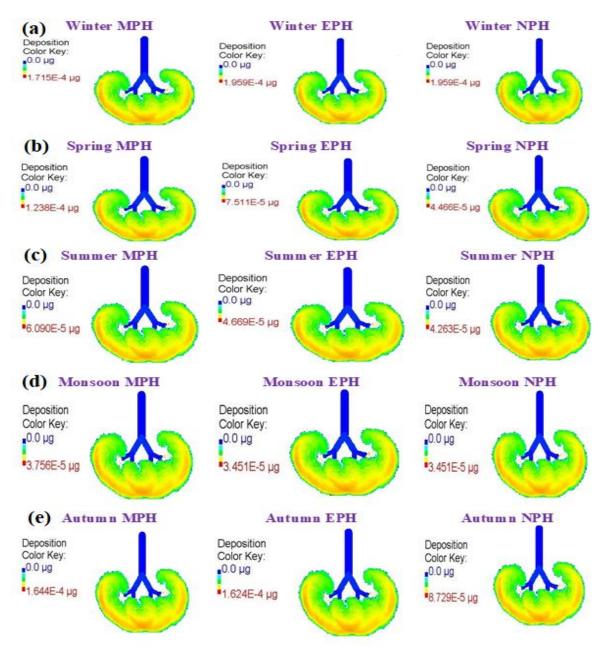


Figure 4.36: Deposition visualization using the mass rate of nanoparticles over Delhi in (a) winter, (b) spring, (c) summer, (d) monsoon, and (e) autumn, respectively. MPH represents morning peak hours, EPH represents evening peak hours, and NPH represents non-peak hours. The deposition rate is plotted as  $\mu$ g/hr and the color scheme represents the deposition density of the particles in human lungs

The deposition fraction of particles in the alveolar region ranges from 1.5 to 2.0  $\mu$ g/min (Fig. 4.36). The order of deposition is alveolar > bronchiole > bronchus at a constant breathing rate of 12 LPM, and the variation in deposition geometry is depicted in Fig. 4.36. This breathing rate is chosen because the breathing rate of an adult is prescribed to be 12 LPM as per the USEPA Handbook, 2011. The breathing rate is kept constant as 12 LPM for all the different age groups to assess the deposition of nanoparticles in the human respiratory system at the same rate. Differences in deposition in different regions occur due to differences in deposition mechanism, particle concentration, and airway geometry (Izhar et al., 2016). The major mechanisms that affect particle deposition in the airways are diffusion, impaction, and sedimentation. The order of deposition ranges from the tracheobronchial (TB) region to the alveolar region based on the size of the particle inhaled. In the TB region, the higher deposition occurs due to the inertial impaction of the particles, and in the bronchiole and alveoli regions, it occurs due to sedimentation (Oliveira et al., 2019). Earlier studies found higher deposition of µm range particles in the head and TB regions with a flow rate ranging from 15 to 30 LPM (Islam et al., 2017).

In the present study, we observe that particles with higher diameters get deposited in the TB region, whereas particles with smaller diameters get deposited in the alveolar region. Higher deposition of particles in the bronchiole region will lead to pharyngitis and rhinitis disease in adults (Islam et al., 2017). The deposition fraction varies during various activities, which alters the breathing frequency. The variation in the deposition of particles in the lungs will subsequently lead to the development of carcinogenic cells based on the rate of deposition of the particles at different regions of the lungs (Gao et al., 2015; Wiebert et al., 2006). For example, adenocarcinoma develops in the outer regions of the lungs, and squamous cell carcinoma in the center region of the lungs (Manojkumar et al., 2019).

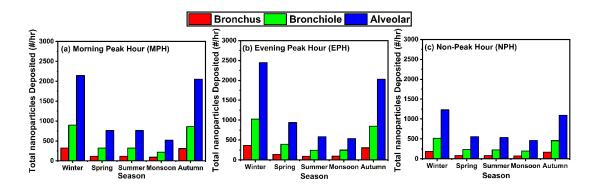


Figure 4.37: Total number of nanoparticles (per h) that get deposited in different parts of the human respiratory system during winter, spring, summer, monsoon, and autumn, respectively, during (a) MPH (morning peak hours), (b) EPH (evening peak hours) and (c) NPH (non-peak hours)

The deposition of nanoparticles in the alveolar region at a constant breathing rate of 12 LPM is highest during EPH of winter season (2445 particles/h and ~6 million particles/season) (Fig. 4.37b). The simulated model values show that the deposition of nanoparticles in different regions during EPH in the winter and autumn seasons is around 5 to 6 million particles, and for the rest of the seasons, it is 4-times lower and is about 1.5 to 2 million particles per season. The MPH deposition ranges from 1 to 6 million in different seasons. During the NPH, nanoparticle deposition in the human respiratory tract ranges from 1 to 3 million. The annual nanoparticle deposition is about 18 million during EPH, 17 million in MPH, and 10 million in NPH, respectively. The IPN predicted by the model based on the particle concentration lies in the range of 0.5 to 1 billion particles for a similar respiratory rate. The simulated model values show that the actual deposition occurring in the human respiratory system is less when compared to the estimated inhalable particle number concentration. The difference occurs due to various other factors that influence the deposition in the lungs. In the study region, the annual average nanoparticle deposition is  $\sim 500 \ \mu g/year$  in the roadside environment; in comparison, the concentration decreased over a location that was away from the roadside environment and was about  $330 \mu g/year$  (Ma et al., 2022). Thus, on an annual scale the exposure concentration in the roadside environment is 30% higher than the environments/regions/locations that are away from roads in urban regions. This increase is attributed to vehicular exhaust over the study region, an urban

locale, which increases the concentration of particles on the roadside compared to an environment away from the roadside. The deposition of particles at a higher rate in the roadside environment poses a major health threat to the people who are working/living in the vicinity of the road (He and Qiu, 2022; Wang et al., 2021). Further, the deposition of nanoparticles is found to be higher in lower respiratory tract infections. These findings from the present study, provide various metrics and indicators for health risk assessment due to urban aerosols/urban pollution. The results also suggest that the inhalation of nanoparticles is more prevalent in the regions where the sources are located, thus, prevention of activities near the sources and source regions of nanoparticles will significantly help to reduce exposure to nanoparticles and associated health impacts.

### 4.6.5 Quantification of nanoparticle deposition in the human respiratory system

The deposition concentrations estimated by the MPPD model are during winter (1.31 x 10<sup>-3</sup>  $\mu$ g/hr) with deposition rates of 0.29  $\mu$ g/hr, followed by spring (1.13 x 10<sup>-3</sup>  $\mu$ g/hr, 0.25  $\mu$ g/hr), summer (8.8 x 10<sup>-4</sup>  $\mu$ g/hr, 0.01  $\mu$ g/hr), monsoon (1.1 x 10<sup>-3</sup>  $\mu$ g/hr, 0.02  $\mu$ g/hr) and autumn (7.88 x 10<sup>-4</sup>  $\mu$ g/hr, 0.01  $\mu$ g/hr), respectively (Fig. 4.38). The deposition concentration estimates are based on the concentration of particles and exposure time. An adult who works around 8 hours a day near roadside conditions experiences a deposition concentration of around 338 µg/year of nanoparticles. Similarly, the adult resident who resides/spends 24 hours a day near the road experiences a deposition of nanoparticles  $>1000 \mu g/year$ . The exposure concentration of particles is 25 times higher than the national annual standards for  $PM_{2.5}$  (40µg/year) and 16 times higher than the  $PM_{10}$  (60µg/year). When compared to the WHO 2021 standard, the exposure concentration is  $\sim 200$  times higher than the PM<sub>2.5</sub> (5µg/year) and 67 times higher than the  $PM_{10}$  (15µg/year) standards, emphasizing the vulnerability of citizens working or living in the vicinity of the road in the study region. It may be noted that the estimated particle deposition only corresponds to nanoparticles and not fine and coarse particles. The exposure estimation shows that the deposition of nanoparticles is found in the head airway tract for children below three years, however, for adults, the deposition is found only in the inner regions of the lungs. The variation in the deposition arises due to the differences in the development of respiratory tract among infants, children and adults. The study region is surrounded by a mixture of educational and residential areas where the population (infants, children and adults) vulnerable to these exposures vary and leads to various health hazards (Balakrishnan et al., 2019; Pandey et al., 2021; Yadav et al., 2022).

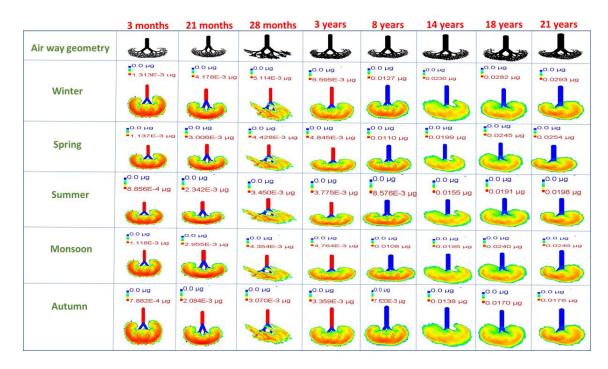


Figure 4.38: Estimated deposition of nanoparticles during different seasons (winter, spring, summer, monsoon, autumn) in the study region of Delhi for age groups from 3 months to 21 years using MPPD model

# 4.6.6 Quantification of total mass deposited during different seasons

The estimated mass of nanoparticles deposited in the respiratory system in a day in the road environment varies between 0.02 and 0.40  $\mu$ g (Fig.4.39). In almost all seasons, the bronchiole deposition is less than 0.05  $\mu$ g/day for different age groups. The alveolar deposition for infants and kids ranges from 0.05  $\mu$ g/day to 0.10  $\mu$ g/day, but for the age groups from 8 years to 21 years, the alveolar deposition is higher, and it varies from 0.30  $\mu$ g/day to 0.40  $\mu$ g/day (Fig.4.39). The estimated alveolar deposition for the 8 to 21 years age group is 0.32-0.40  $\mu$ g/day in winter, 0.25-0.32  $\mu$ g/day in spring, 0.20-0.27  $\mu$ g/day in summer, 0.25-0.33  $\mu$ g/day in monsoon and 0.20-0.25  $\mu$ g/day in autumn. In winter, the deposition concentration is 2 times higher than in autumn for adults. The average nanoparticle deposition in the alveolar region of a

particular individual, irrespective of the age groups residing near the road in different seasons is 26.6  $\mu$ g/day in winter, 14.6  $\mu$ g/day in spring, 22  $\mu$ g/day in summer, 23.1  $\mu$ g/day in monsoon and 17.1  $\mu$ g/day in autumn. The seasonal order of deposition is winter > monsoon> summer > autumn > spring.

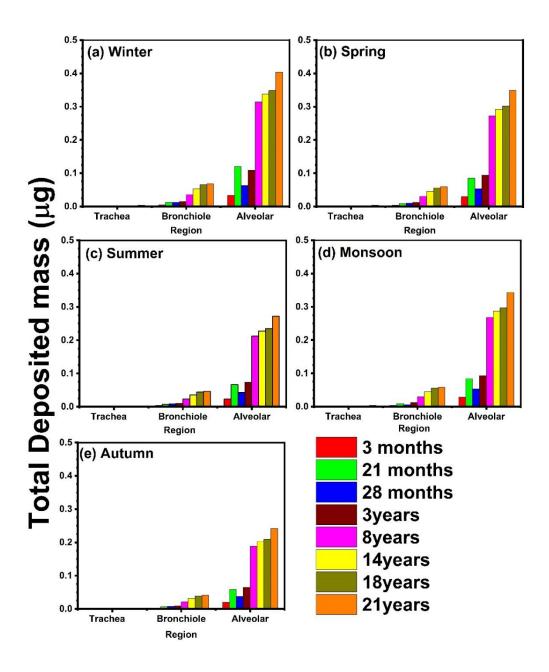


Figure 4.39: Estimated total deposited mass in different regions of respiratory system (trachea, bronchiole, and alveolar) during (a) winter, (b) spring, (c) summer, (d)

monsoon and (e) autumn in the study region of Delhi for different age groups ranging from from 3 months to 21 years using MPPD model

The seasons with more number of days in them account for more exposure periods and vice versa. It is estimated that in a year atleast about 100  $\mu$ g of nanoparticles are deposited in the alveolar regions alone in adults in the study region. The age wise order of deposition of nanoparticles is adult (18 years and above) > adolescent (14 years) > child (3 to 8 years) > toddler > (21, 28 months) > infant (3 months). Thus, it is clear that, in urban areas where most residents are located near the road the vulnerability to nanoparticles is higher compared to other environmental regimes (rural, semi-ruban) irrespective of the exposure time or specified age group.

# 4.6.7 Quantification of total particle numbers deposited in different seasons

The daily deposition of nanoparticles in the alveolar regions ranges between 5 and 8 x  $10^5$  for adults in different seasons (Fig.4.40). The deposition of nanoparticles is estimated to be less in infants (3 months), gradually increases in children, and becomes higher in adults. Less than 2% of particles are deposited in the trachea region and 6-8% in the bronchiole region, whereas the remaining 90% get deposited in the alveolar regions. Due to its physical properties, most particles get deposited in the alveolar regions. Due to lesser mass, the nanoparticles penetrate the head and TB regions, so more deposition is found in the alveolar regions (Viitanen et al., 2017). The estimated total nanoparticle deposition in different regions of the human respiratory tract varies based on respiratory parameters such as tidal volume, functional breathing capacity, and breathing frequency. The age group of 21 years and above experiences more nanoparticle deposition than other age groups (Fig. 4.40). Nanoparticle deposition in the human respiratory tract can cause various health issues such as asthma, leukemia, prostate, and olfactory disorders (Manigrasso and Avino, 2012), thus, necessary mitigation measures need to be taken to reduce nanoparticle exposure, especially in urban regions.

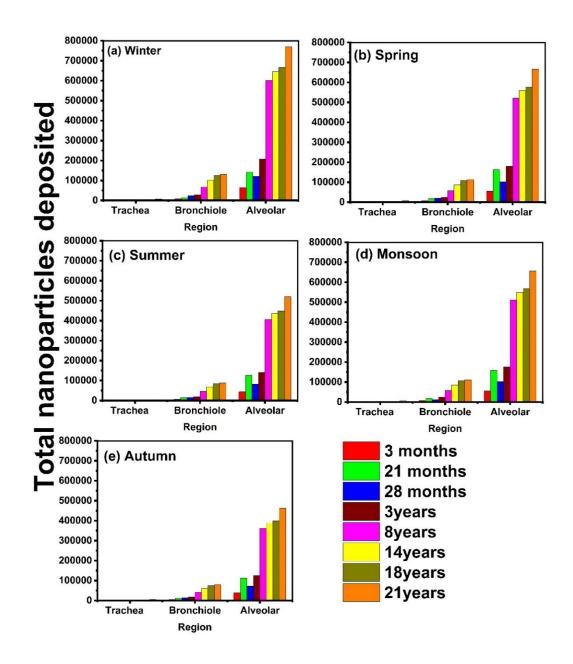


Figure 4.40: Total nanoparticle deposition estimated using MPPD model for different age groups in different regions of respiratory system, namely, trachea, bronchiole, and alveolar in (a) winter, (b) spring, (c) summer, (d) monsoon and (e) autumn) over Delhi

# 4.6.8 Deposition fraction of particles in different regions in the human respiratory tract

Deposition fraction indicates the fraction of nanoparticles inhaled which then get deposited in the respiratory system. The factors that influence the deposition include the mechanism and the airway structure which are based on the mass median diameter of the aerodynamic particles. The analysis shows that the deposition fraction in trachea region is minimal (<0.01). In the bronchiole region the deposition fraction is almost the same (~0.03) in all the seasons, suggesting that the deposition in the bronchiole region is less influenced by the seasonal variation in concentration and meteorology. The deposition fraction in the alveolar regions varies based on age and seasonal concentration.

The deposition fraction varies between 0.02 and 0.04 in the alveolar regions for different seasons. In winter, the fraction exceeds 0.04 for infants, suggesting that infants are more vulnerable during winter. The deposition in the alveolar region is 4-5 times greater than the deposition in the bronchiole region (Fig. 4.41). The difference in deposition in alveolar regions for adults is around 30 to 40% higher compared to infants and children. A previous PM<sub>2.5</sub> concentration-based deposition study also reported that the total deposition of particulate matter is seen more in children compared to other age groups (Manojkumar and Srimuruganandam, 2022) consistent with the present study. Several other studies showed that the particles with higher diameter deposit in the head region, and those with lower diameter deposit in the alveolar regions, which are corroborated by the present study.

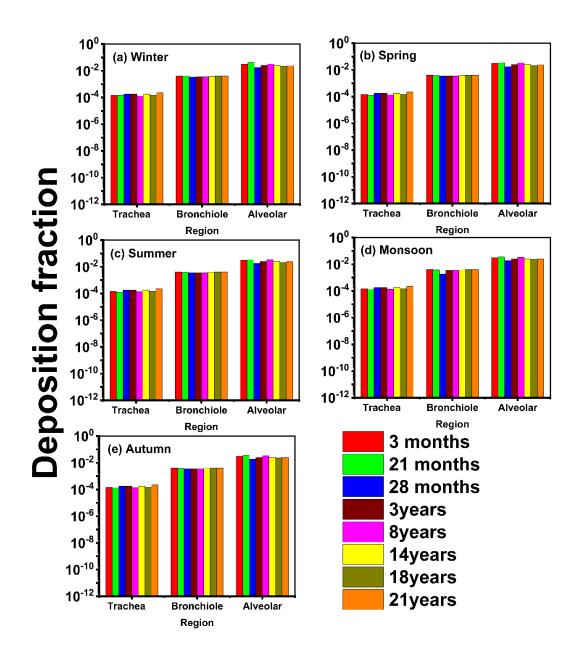


Figure 4.41: Deposition fraction of nanoparticles estimated using MPPD model for different age groups in the trachea, bronchiole, and alveolar regions of respiratory system during (a) winter, (b) spring, (c) summer, (d) monsoon and (e) autumn in the urban roadside study region of Delhi

## 4.7 Nanoparticle concentration analysis during episodic event

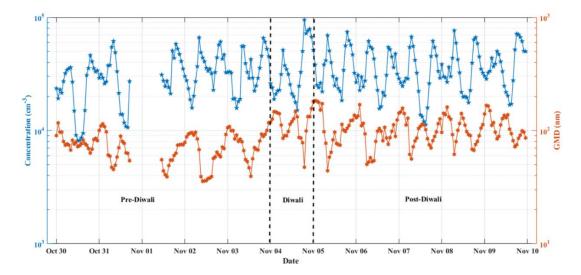
#### 4.7.1. Firework episodic event

Diwali is one of the major festivals celebrated annually in India. The festival is celebrated in almost all parts of the country. It is celebrated by the bursting of crackers and lighting lamps (Ganguly, 2015; Garg and Gupta, 2018; Yadav et al., 2022a; Yadav et al., 2019). The crackers used for firework contains a large number of chemicals such as aluminum, sulfur, potassium nitrate, barium nitrate, charcoal, manganese, strontium nitrate, potassium and iron dust powder as a composition material for the manufacturing of firecrackers (Nishanth et al., 2012; Perrino et al., 2011; Sateesh et al., 2018). The lighting of fireworks emits different types of emissions into the atmosphere. It releases particulate matter, gaseous pollutants and toxic metals of significant quality. These particles stay in the atmosphere for a few days causing the formation of toxic smog. The pollutants are hazardous in nature and cause serious health effects as well (Kanawade et al., 2014). The study location is the national capital of India, New Delhi and is already facing severe pollution events throughout the year and majority of the time, the air quality remains in the poor category (Agarwal et al., 2020; Mishra et al., 2016; Mohan and Mishra, 2022). The air quality standards exceed national ambient air quality standards (Garg and Gupta, 2018; Goyal et al., 2021; Kanawade et al., 2020). The event Diwali falls under the post-monsoon season (beginning of the winter season) from October to November every year (Garg and Gupta, 2018; Ghei and Sane, 2018). During winter, air quality becomes poor due to the prevailing meteorological conditions such as relative humidity, boundary layer height and ventilation (Gani et al., 2020, 2021; Ramachandran and Rupakheti, 2022). Further increased emissions on Diwali also add more pollutants to the atmosphere. Recent studies found that air quality worsens during Diwali (Chatterjee et al., 2013; Kanawade et al., 2020; Parkhi et al., 2016; Yadav et al., 2022a; Yadav et al., 2019). People also call Delhi a Gas chamber during the pollution event. Apart from Diwali, Delhi is facing emission issues from sources such as vehicles, construction sites, road dust emission, industry sector, waste burning and also due to long-range transportation of pollutants of different geographical origins (Gani et al., 2019, 2020; Kumar et al., 2015; Mishra et al., 2016). Several studies in Delhi analyzed the impacts of crackers

on the local air quality and only very few studies reported an analysis of particles in number concentration (Garg and Gupta, 2018; Parkhi et al., 2016; Sateesh et al., 2018; Yadav et al., 2022b). This study analyzed the concentration and temporal variation of nano-size particles ranging from 10 to 1090 nm. The wide-range nanoparticles are classified into four different sub-categories such as Nucleation (10 to 30 nm), small Aitken (30 to 50 nm), large Aitken (50 to 100nm) and Accumulation mode (100 to 1000nm). The study focuses on the size distribution of different size range particles during the event of Diwali and the period adjacent to it. The study helps in understanding the pattern of emission of nanoparticles during the firework event. Because of the restriction on firecrackers use in the city, a reduction is found in the concentration of the particles. Although the use of fire crackers is restricted in the city but few fire work events still occurred on Diwali evening as a part of the tradition.

## 4.7.2 Temporal variation of Particle number concentration during episode

The hourly average concentration of the total particle number concentration (PNC) of the nanoparticles was found high during Diwali, especially during the evening hours when the fireworks event occurs regularly. The remaining days in the pre- and post-Diwali days represent the emission from the transportation sector that shows two peaks in a day during morning and evening peak hours. The concentration of the Diwali day reaches a maximum of about 9 x  $10^4$  cm<sup>-3</sup> (Fig. 4.42) whereas the maximum concentration during the pre and post-Diwali was found in the range of 6 to  $7 \ge 10^4$  $cm^{-3}$  (Fig. 4.42). During Diwali, the concentration of PNC was increased about 30% comparing pre-Diwali period and 19 % comparing post-Diwali period. The change in the percentage of concentration during the pre and post-Diwali periods was due to the existence of particles from the Diwali emissions especially in the post-Diwali period. The geometric mean diameter of the particle also showed an increase in their mean diameter during the Diwali period and the maximum concentration of the GMD was obtained during the post-Diwali phase compared to other phases due to the coagulation of the particles after the Diwali day. The different size particles such as N<sub>nuc</sub> and N<sub>satk</sub> were seen in less concentration during the Diwali period since they were not directly associated with Diwali emissions, Nlatk (50 to 100 nm) was slightly associated with Diwali emissions and Nacc was observed in higher concentration showed a majority of the particle was found above 100 nm from the Diwali emissions. The statistical analysis of different size particles during different monitoring periods with their minimum and maximum concentration was shown in Table.4.9.



*Figure 4.42: Daily variation of hourly averaged particle number concentration and GMD from 30<sup>th</sup> October to 10<sup>th</sup> November 2022* 

# 4.7.3 Size resolved particle distribution

The size distribution of the different size-resolved particles is plotted for three periods, namely the pre, post and Diwali periods, for distinguishing the different peaks occurring at different periods. The size distribution of the particles shows a clear single peak of higher concentration during the Diwali day in the size range of particles around 150 nm in the accumulation mode of the particles. The pre and post-Diwali concentrations show a gradual rise in the concentration from smaller to larger size, attaining a peak in the accumulation mode, whereas on Diwali, the N<sub>nuc</sub> concentration decreases to a certain extent and then it starts to increase gradually from N<sub>satk</sub> to N<sub>acc</sub> (Fig. 4.43). This shows the presence of less particles during the period compared to another period. The concentration of the particle pattern was totally different from N<sub>nuc</sub> to N<sub>latk</sub> and N<sub>acc</sub>. During the pre-Diwali period, the lower size particle till N<sub>latk</sub> mode concentration was found to be high, followed by the post-Diwali period and then by the Diwali period (Fig. 4.43). In N<sub>acc</sub> the pattern was totally opposite i.e., higher concentration on Diwali period followed by a post-Diwali period and then by Diwali period. This indicates a clear representation of the particles and their sources. The

smaller size particle till 100 nm is directly associated with vehicular emissions found in the Pre and Post-Diwali period. During Diwali period, particles above 100 nm are found in higher concentrations and are associated with Diwali emissions.

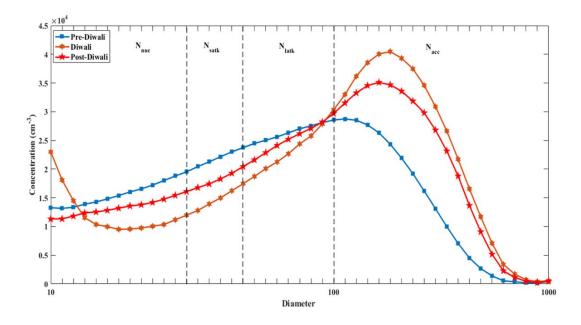


Figure 4.43: Particle number size distribution of  $N_{nuc}$ ,  $N_{satk}$ ,  $N_{latk}$  and  $N_{acc}$  during  $30^{th}$ October to  $10^{th}$  November 2022

## 4.7.4 Particle composition in total number concentration

The composition of different size particles throughout the monitoring period was done to analyze the role of different size particles in the total number concentration. This shows the percentage contribution of different size particles. The contribution of smaller size particles in total concentration was found to be high during the pre-Diwali period. The accumulation mode alone contributed to a higher extent of around 57% (Fig. 4.44) of the total particles on the Diwali day, followed by 48% to 55% in the post-Diwali period. The contribution of the fire emissions during the Diwali and post-Diwali was seen clearly in the figure 4.44. The total PNC concentration is based on the contribution from different size fractions emitted from different sources such as engine exhaust, biomass burning, and fire emissions in the urban regions. The hourly contribution of the different sizes to the total PNC is also analysed to clearly understand the emissions. The hourly emission contribution of the different size particles to the total PNC shows on Diwali day. The contribution of the accumulation mode dominates 60 to 70% of the PNC during the Diwali evening and the next morning (Fig. 4.45b).

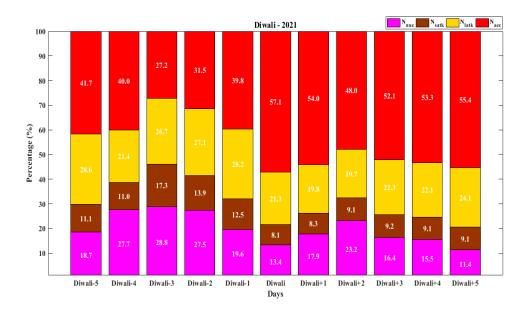


Figure 4.44: Percentage contribution of  $N_{nuc}$ ,  $N_{satk}$ ,  $N_{latk}$  and  $N_{acc}$  in Total PNC (30<sup>th</sup> October to 10<sup>th</sup> November 2022)

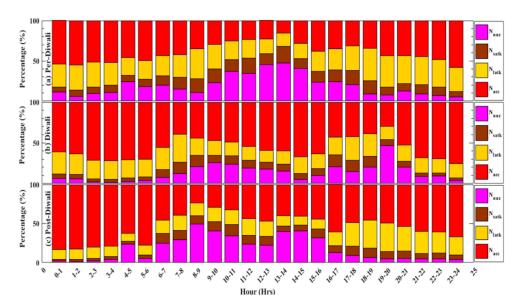
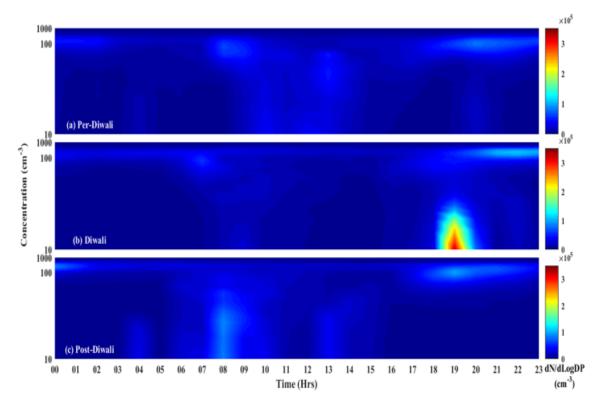


Figure 4.45: Hourly classified percentage contribution of N<sub>nuc</sub>, N<sub>satk</sub>, N<sub>latk</sub> and N<sub>acc</sub> in Total PNC during (a) pre-Diwali, (b) Diwali (c) post-Diwali from 3<sup>rd</sup> November to 5<sup>th</sup> November 2022

The Pre Diwali period showed an increase in the concentration of the Nucleation mode particles during the afternoon hour due to the effect of sun rays which causes nucleation burst events (Kompalli et al., 2018). Since the monitoring period falls under winter seasons, the higher concentration of solar radiation and ventilation was seen during the afternoon in Delhi (Fig. 4.45a). This can be clearly visualized during the Post Diwali day when the solar radiation starts to increase around 06:00 hrs, the concentration of the nucleation mode particles also seen increased (Fig. 4.45c). Whereas during the off-sunshine hours the nucleation mode particles again start to decrease and the concentration of PNC was dominated by the accumulation mode particles due to coagulation.



### 4.7.5. Heat map analysis during episodes

*Figure 4.46: Heat map analysis of size-resolved particle distribution and GMD during* (*a*) *pre-Diwali*, (*b*) *Diwali and* (*c*) *post-Diwali from 3<sup>rd</sup> November to 5<sup>th</sup> November* 2022

The heat map analysis is used to evaluate different size particles with respect to time. During Pre-Diwali period, in early morning and late evening, a thin hazy layer of particles around 100 nm was found (Fig. 4.46a) in the concentration range of 1 x  $10^5$  cm<sup>-3</sup> due to the accumulation of particles under low boundary layer conditions and less dispersion of particles. During the Diwali period in the evening around 18 to 20 Hrs. There is a clear visibility of particle emissions from fireworks in the range of 3 to 2 x  $10^5$  cm<sup>-3</sup> (Fig. 4.46b) and the hazy layer of particles in pre-Diwali is seen in a similar pattern in the late evening and early morning on Diwali day with higher concentration due to the pollution load increased by Diwali. During the post-Diwali phase, during the morning (8 Hrs.), another emission source of around 1 x  $10^5$  cm<sup>-3</sup> was due to the vehicular emissions, which added more particles apart from preexisting Diwali emission particles. In the Pre Diwali period, the vehicular emission contribution during the morning and evening peak hours is also clearly visible.

#### 4.7.6 Correlation analysis of different size particles

The correlation analysis of particles with gaseous pollutants shows that the  $N_{acc}$  and  $N_{latk}$  show moderate to good correlation with particles associated with Diwali emissions and the correlation values are found to be higher during the post-Diwali phase. Certain pollutant, such as O<sub>3</sub> showed very less correlation (r<sup>2</sup> =0.1 to 0.2) (Fig. 4.47c) throughout the Diwali phases since O<sub>3</sub> was not directly associated with Diwali emissions.

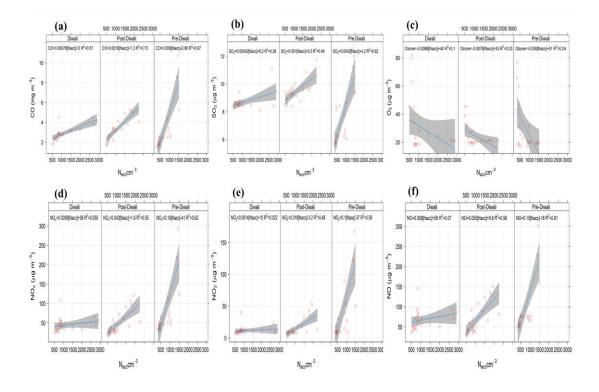


Figure 4.47: Correlation analysis of  $N_{acc}$  with(a) CO (b)SO<sub>2</sub> (c)O<sub>3</sub> (d)NO<sub>x</sub> (e)NO<sub>2</sub> (f)NO during pre, post and Diwali period from 30<sup>th</sup> October to 10<sup>th</sup> November 2022

The Nitrogen components such as NOx, NO<sub>2</sub>, NO showed the highest correlation values. The  $r^2$  values ranged from 0.4 to 0.6 (Fig. 4.47 d-f) for the N<sub>acc</sub> mode particles and for particles in N<sub>latk</sub> mode the correlation values was found to be high in the post Diwali phase ( $r^2 = 0.7$ ) compared to the pre-Diwali ( $r^2 = 0.2$  to 0.3) and Diwali phase ( $r^2 = 0.01$  to 0.02) (Fig. 4.48 d-f). The  $r^2$  of CO ranges from 0.6 to 0.7 (Fig. 4.47a) for the N<sub>acc</sub> and for N<sub>latk</sub> mode ( $r^2 = 0.2$  to 0.4) (Fig. 4.48a). The carbon emissions were high before Diwali and on Diwali, the concentration was found less and then during the post-Diwali phase, the values started to resume back to its original concentration. The correlation values were insignificant for the smaller sizes, such as N<sub>nuc</sub> and N<sub>satk</sub> mode particles.

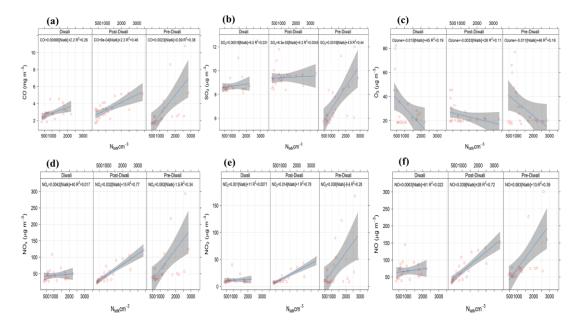


Figure 4.48: Correlation analysis of  $Nl_{atk}$  with(a) CO (b)SO<sub>2</sub> (c)O<sub>3</sub> (d)NO<sub>x</sub> (e)NO<sub>2</sub> (f)NO during pre, post and Diwali period from 30<sup>th</sup> October to 10<sup>th</sup> November 2022

### 4.7.7 Summary

The particle number concentration of the nano particles ranging from 10 to 1090 nm was measured from  $30^{\text{th}}$  October to  $10^{\text{th}}$  November 2022. This period covers the major festival of the country like Diwali, which is usually celebrated through bursting of firecrackers. The Diwali day received the maximum particle number concentration of about 9 x  $10^4$  cm<sup>-3</sup>. In the previous period and after the Diwali period, concentrations ranged from 6 to 7 x  $10^4$  cm<sup>-3</sup>. On Diwali day, a 30 % increase in the concentration of the particles was observed compared to the normal day emissions. The increase in emissions was due to the fireworks events that occurred. In the post-Diwali period, the concentration of Diwali emissions along with normal emissions. The particles in size range > 100 nm i.e., N<sub>acc</sub> mode particles were seen in higher concentration. The size distribution of the particles showed a peak concentration of particles in the range of Nl<sub>atk</sub>, i.e., 50 to 100 nm size particles were seen in peak concentration.

Table 4.9: statistical summary of different size pollutants during the different phases of the episodic event

S.no	Parameter	Value	Pre Diwali	Diwali	Post Diwali
1.	N <sub>nuc</sub> (cm- <sup>3</sup> )	Minimum	87.3	27.4	38.8
		Maximum	2062.5	3428.2	2646.1
		Mean $\pm$ SD	$606.2 \pm 462.2$	473.1 ±689	$514 \pm 495.8$
2.	N <sub>satk</sub> (cm- <sup>3</sup> )	Minimum	149.6	136.1	106
		Maximum	2455.9	1429.1	2336.3
		Mean $\pm$ SD	878.8 ± 539.3	599.2 ±396	732.5 ±551
3.	Nlatk (cm- <sup>3</sup> )	Minimum	284.0	339.2	318.2
		Maximum	2842.7	2233.7	3483.2
		Mean $\pm$ SD	$1113.1 \pm 674.4$	$1003 \pm 572$	$1074 \pm 781$
4.	Nacc (cm- <sup>3</sup> )	Minimum	158.7	506.1	317.8
		Maximum	1505.7	2902.1	2258.1
		Mean $\pm$ SD	$598.4\pm278$	1111 ±714	$961.4 \pm 337$
5.	Ntotal (cm- <sup>3</sup> )	Minimum	8156	15120	11263
		Maximum	66314	95073	76588
		Mean $\pm$ SD	$32550 \pm 13957$	38289±23038	37203±15796
6.	GMD (nm)	Minimum	35	47	44
		Maximum	117	156	183
		Mean $\pm$ SD	73±20	119±27	104±31
7.	NO <sub>2</sub> ( $\mu$ g/m <sup>3</sup> )	Minimum	6.80	6.6	6
		Maximum	187.2	40.2	68.8
		Mean $\pm$ SD	$24.9 \pm 31.7$	11.7 ±7	$13.1 \pm 11.4$
8.	NO ( $\mu g/m^3$ )	Minimum	24.6	37.3	30.8
		Maximum	319	145.4	200.4
		Mean $\pm$ SD	$76.7 \pm 45.5$	67.6 ±24.2	77.2 ±39.1
9.	NOx ( $\mu g/m^3$ )	Minimum	18.5	25.3	21.5
		Maximum	318.5	108.5	160.7
		Mean $\pm$ SD	$59.7 \pm 47.4$	44.4 ±17.9	50.61 ±28.3
10.	SO <sub>2</sub> ( $\mu g/m^{3}$ )	Minimum	5.1	8.1	8
		Maximum	12.5	11	11.8
		Mean $\pm$ SD	$6.2 \pm 1.2$	8.6 ±0.5	8.7 ±0.5
11.	CO (mg/m <sup>3</sup> )	Minimum	0.3	1.8	0.9
		Maximum	11.4	4.5	5.6
		Mean $\pm$ SD	$2.3 \pm 1.81$	$2.8 \pm 0.7$	2.7 ±0.9
12.	Ozone (µg/m <sup>3</sup> )	Minimum	17	18	18.5
		Maximum	111.3	82	101.4
		Mean $\pm$ SD	37.1 ± 25.1	30.6 ±19.1	32.9 ±22

These size ranges are directly associated with exhaust emissions in urban regions.  $N_{acc}$  mode particles were observed in higher concentrations during Diwali days and the concentration was about 57% of the total concentrations, whereas before Diwali, the contribution was around 31 to 40 %. During the evening hours, when the firework events occured the percentage contribution of  $N_{acc}$  mode particles to the total

concentration reached around 60 to 70 %. During the pre-Diwali period, N<sub>nuc</sub> particles in the mid-day received a higher concentration of about 25 to 28% due to the burst of particles. The heat map analysis showed clear evidence of a hazy thin layer of particle emissions during the morning and evening peak hours resulting from vehicular emissions. During Diwali day, especially during 18 to 20 Hrs. The particles' concentration was around 2 to 3 x  $10^5$  cm<sup>-3</sup>. The correlation of gaseous pollutants shows a good correlation with N<sub>acc</sub> and Nl<sub>atk</sub> mode particles and for other smaller sizes, the correlation was not found. NOx, NO<sub>2</sub>, NO showed a good correlation with N<sub>acc</sub> (r<sup>2</sup> =0.4 to 0.6). The highest correlation of NOx, NO<sub>2</sub>, NO that found throughout the monitoring period was (r<sup>2</sup> = 0.7) with N<sub>latk</sub> mode particles during the post-Diwali phase. The correlation of N<sub>acc</sub> mode with CO was observed to be high during the pre-Diwali period (r<sup>2</sup> =0.6 to 0.7) and less during the Diwali period. The study revealed the role of Diwali festival emissions on Particle number concentration of nano pollutants in the atmosphere. However, the usage of firecrackers in the National capital Delhi is reducing year to year due to the air pollution events occurring after Diwali.

# CHAPTER – 5 CONCLUSIONS

### 5.1 Introduction

Deteriorating air quality in Delhi, India, has created global interest in understanding its issues due to its complex mixture of sources. The study analyzed the variation of different size nano pollutants in the region ranging from 10 to 1090 nm for a year period from December 2020 to November 2021. The study period includes two major classifications one is based on the changes in emission sources intensity and another one is based on the seasonal variation. The first part was conducted in the roadside environment near a busy road in Delhi from 1st April 2021 to 30th June 2021 (classified as Period I), and from 3rd October 2021 to 30th November 2021 (referred to as Period II) based on emission sources variation. The seasonal variation (winter, spring, summer, monsoon, and autumn) was analysed in megacity city Delhi for the first time. The nano-sized pollutants exhibit significant diurnal and seasonal variations due to differences in environmental conditions, emission sources, regional meteorology, and atmospheric dynamics. So, in this study, deposition of nanoparticles (size ranging from 10 to 1000 nm) in the human respiratory system using a Multiple Path particle Dosimetry (MPPD) over the urban megalopolis Delhi is estimated. The size distribution of particles in an urban roadside environment in the above size range is examined with respect to vehicular flow density, such as during the peak (morning, and evening) and non-peak hours. The major conclusions of the study are as follows.

# 5.2 Summary of nanoparticles concentration and meteorological role during different emission scenarios

The study found that the concentration of the particles in urban roadside microenvironments varies with increased or reduced anthropogenic activities, especially vehicular-based emissions. During Period I (during restriction (DR) phase), due to the various restriction for vehicular movement and other anthropogenic sources, less concentration was recorded in the study area  $(1.7 \times 10^3)$ . During the same DR phase, the recorded vehicle fleet was 49% less than the normal period. Similarly, after Diwali (AD) phase, which occurs immediately after Diwali, a maximum concentration

of particles  $(1.3 \times 10^4 \text{ cm}^{-3})$  was recorded. This was due to the contribution of both vehicular emissions and Diwali firecrackers emissions. The PNC concentration in AD phase increased by 35% compared to the previous phase. The particle number concentration found in the range of 10<sup>3</sup> cm<sup>-3</sup> to 10<sup>5</sup> cm<sup>-3</sup> during Periods I and II, except for the AD phase in Period II, in which PNC ranged from 10<sup>4</sup> cm-3 to 10<sup>5</sup> cm<sup>-3</sup>. Apart from PNC, the geometric mean diameter (GMD) and size distribution of the particles also vary, which help to identify the source contributors. When emissions are complex, the GMD of the particles varies from 20 nm to 200 nm (Period II), but when emission sources are restricted, the range also narrows to 15 to 80 nm (Period I). Periods I and II received average precipitation of about 116 mm and 80 mm, respectively, resulting in the wet removal of particles in large quantities. In the entire study period, Period I experienced around 31 % less concentration of particles (~2.4 x 10<sup>4</sup> cm<sup>-3</sup>) due to lockdown restrictions and, on the other hand, particle concentration was increased by 35% compared to normal conditions due to the sudden rise in firework emissions in Period II. In addition, the size distribution of particles also changed between two periods. The contribution of different size bins to the total concentration varies according to the sources. During the DR phase, when there was a restriction, the contribution of N<sub>nuc</sub> and N<sub>satk</sub> was less to N<sub>total</sub> since they were directly associated with the vehicle-related exhaust in urban environment. N<sub>nuc</sub> contribution is high in BR and DR phases, which are marked by average emissions. Events like Diwali and pollution events add more particles into the atmosphere along with the pre-existing regular emissions over an urban region, so the contribution of Nacc in Ntotal was high during the AD and DR II phases. Local and regional meteorology also played a significant role in deciding the PNC in the roadside environment. When the relative humidity was high, the coagulation of the particles occurs to a greater extent, and hence, the concentration becomes high (Period II). The diurnal behaviour showed high concentrations of the pollutants during the peak hours (morning and evening) because of vehicular emissions during normal conditions (without restriction). Wind speed and direction were the other factors influencing the concentration of these particles. Dispersion of particles occurred based on wind speed and direction: higher wind speeds resulted in higher dispersion, and vice versa.

The study presents detailed characteristics of PNC in the urban city of Delhi during various emission scenarios. The study confirms the role of traffic emissions in metropolitan cities in determining the particle load. The PNC estimates will be useful to determine the deposition of particles in the human respiratory system based on the various inhalation rates and associated physical activities. The particles in the nanosize range after deposition can potentially transport to other parts of the human body, creating more chronic and acute illnesses in the human body. The quantitative outcomes of the present study can thus be used to estimate human health impacts, develop policies/standards, and initiate mitigation measures for pollution events with implications to climate change and help move towards sustainability measures.

### 5.3 Summary of nanoparticles concentration and during different seasons

The nano-sized pollutants exhibit significant diurnal and seasonal variations due to differences in environmental conditions, emission sources, regional meteorology, and atmospheric dynamics. The concentration of particles was classified into four different sizes as N<sub>nuc</sub> (10 to 30 nm, nucleation particles), N<sub>satk</sub> (30 to 50 nm, small Aitken particles), Nlatk (50 to 100 nm, large Aitken particles), and Nacc (100 to 1000 nm, accumulation mode particles), and the total particle number concentration (PNC) as N<sub>total</sub> (or total PNC). PNC ranged between 10<sup>4</sup> cm<sup>-3</sup> and 10<sup>6</sup> cm<sup>-3</sup> over Delhi during the year, and the highest concentration is found during winter. On a seasonal scale, the concentration in winter was  $\sim 2$  times higher than in monsoon, 1.6 times higher than in summer, and about 1.3 times higher than in autumn and spring. The diurnal variation analysis showed distinct characteristics of seasonal, primary, and secondary emissions throughout the day. The direct primary emissions from engine exhaust exhibit a prominent double hump structure during morning and evening peak hours in winter and autumn. The double hump structure was same for PNC and other gaseous pollutants related to vehicular emissions during these seasons. Appropriate correlation was found between carbon monoxide (CO) and  $N_{acc}$  (100 to 1000 nm) in all seasons ( $R^2 \sim 0.5$ ), suggesting that emissions from fossil fuel combustion was the predominant source for gases and particles in the near-road curbside environment. Ozone concentration increaseed multi-fold during the peak sunshine hours due to photochemical reactions with nitrogen oxides (NOx) and volatile organic compounds (VOCs) under the influence of sunlight and heat in all the seasons except in monsoon. The study also revealed that apart from differences in primary direct emissions, the particles undergo secondary transformation (through coagulation and condensation) due to the prevalent meteorological conditions that differ in different seasons.

The concentrations in different size ranged (N<sub>nuc</sub>, N<sub>satk</sub>, N<sub>latk</sub> and N<sub>acc</sub>) and their respective contributions to total PNC exhibit significant seasonal variations. During winter, larger size particles (large Aitken and accumulation) contributed more to the total due to coagulation, with accumulation mode alone contributing >40% to the total PNC. In winter, the concentration of accumulation mode particles (>500 nm) is higher in the evening hours due to the coagulation of smaller particles. N<sub>nuc</sub>, N<sub>satk</sub>, and N<sub>latk</sub> (<100 nm) were higher in spring and summer during mid-day due to nucleation and/or ultrafine particle (UFP) burst events. The concentrations in the nucleation mode (smaller size) and accumulation (larger size) exhibited higher deviations in their respective contributions to the total, with N<sub>nuc</sub> contributing 20% to 36% and N<sub>acc</sub> accounting for 23% to 45% in all the seasons. In contrast, the concentrations in the medium size exhibit lesser variability as the contribution of Nsatk varied to Ntotal from 12% to 17%, and N<sub>latk</sub> varied from 22% to 26%, respectively during the year. The particle number concentrations and their contributions were influenced by differences in different emission sources and due to variations in other meteorological parameters such as wind speed, relative humidity, temperature, solar radiation and boundary layer height during day and night. The percentage contributions by different size particles to N<sub>total</sub> during daytime and nighttime varied and the results clearly indicate the distinct role of diurnal variation in emission sources. The percentage contributions of nucleation mode followed by accumulation contribute >60% to N<sub>total</sub> during daytime with their sum contribution being the highest in spring season (67%). During the warm periods of spring, summer and monsoon, Nnuc contributes ~45% to Ntotal during daytime due to the prevalent meteorology conditions which enabled freshly emitted exhaust emissions to remain in the atmosphere longer. Whereas in autumn and winter N<sub>nuc</sub> contributed <30% to N<sub>total</sub> in daytime. In stark contrast, the concentration of large particles, Nacc and Nlatk, dominated N<sub>total</sub> during nighttime, and in particular in winter and autumn Nacc and Nlatk contributed 77% to Ntotal owing to coagulation of emitted

particles as the prevalent atmospheric conditions (higher RH, lower temperature and BLH) favoured coagulation, with  $N_{acc}$  alone contributing ~50% in both the seasons. The contribution of  $N_{satk}$  and  $N_{latk}$  was less to  $N_{total}$  during daytime and nighttime and they do not exhibit significant seasonal variabilities –  $N_{satk}$  contributes 14-17% (daytime) and 10-16% (nighttime, whereas  $N_{latk}$  contributes 18-25% (daytime) and 23-29% (nighttime), respectively. The size distribution of particles in an urban roadside environment in the above size range is examined with respect to vehicular flow density, such as during the peak (morning, and evening) and non-peak hours. The UFP contributed 60 (autumn) to 80% (monsoon) to the total particle concentration in the study region. The concentration of UFP and  $N_{acc}$  during non-peak hours was less than that of peak hours, confirming the dominant influence of emissions from vehicular exhaust in the study region. The size distribution analysis of UFP in different seasons clearly identified the role of various emission sources such as direct emissions (winter, and autumn) and particle formation process (spring, and summer).

This study provides insights into the behavior, dynamics, contribution, and characteristics of air pollutants in the nano-size range in the atmosphere. The study concludes that the dynamics and profiling of nanoparticles on the environment and human health are not only related to the concentration of direct primary emissions but also depend on the gaseous precursors, meteorological conditions, and atmospheric dynamics of/and over a particular region. The results provide quantitative insights into the number concentrations of particles (air pollutants) in different size ranges from nucleation to accumulation and their behavior in a roadside urban environment of a highly polluted urban city on seasonal scales, which will be useful in devising control strategies aimed to improve air quality, public health, environment, and climate. Concentrations of particles from nucleation to accumulation modes are significantly higher in Delhi in winter than other seasons. Air quality across north India is found to worsen during winter due to increases in aerosols, and rising relative humidity (Paulot et al., 2022). Further, due to rapid industrialization and growth in population improving air quality is seen to be a big challenge for India (Xie et al., 2024). The winter PM<sub>2.5</sub> concentrations decreased in 2022 by 20%, out of which one-half of the reduction was attributed to favorable meteorological conditions in the last 5 years, which are unlikely to persist in a warming world (Xie et al., 2024). An estimated 1.7 million premature mortalities in 2019 in India were attributed to severe surface-level air pollution according to a Global Burden of Disease study. Further, an estimated 15% of deaths in South Asia were linked to health effects of air pollution exposure (World Air Quality Report, 2020). These quantitative results on seasonal variations of air pollutants together with the knowledge on seasonal variations in meteorological parameters and atmospheric dynamics (e.g., lower BLH in winter favors trapping the pollutants near surface as opposed to summer) form a basis which can be effectively utilized while devising mitigation measures to enhance the positive effect of improving the air quality and public health.

# 5.4 Role of regional precursors and meteorology in determining the nanoparticles concentration in the study area

The nanoparticles concentration was not only associated with the intensity of the sources but also with local factors such as gaseous emissions in the region and meteorological conditions. Vehicular emissions contributed to nanoparticle concentration and other co-pollutants, such as NO, NO<sub>2</sub>, NO<sub>x</sub>, SO<sub>2</sub>, CO, BC, and O<sub>3</sub>. The evolution of the gaseous pollutant's concentration had similar trends to particles because they have the same origin (traffic emissions). The BC used as a tracer for transport emissions showed higher concentrations during nighttime. The concentration of gaseous pollutants also increased to around 3 to 5% during winter. Winter season in the study period experienced a shallow boundary layer < 1000 m, which promoted the pollution accumulation at the ground level, while during summer, the boundary layer reached up to 2400 m, which helped greater dispersion. Based on the wind speed and boundary layer height, the ventilation coefficient of the region also changed seasonally. Precipitation was another important factor for determining a region's total particle number concentration because it helped in wet scavenging of the pollutants, especially in the N<sub>acc</sub> mode particles. The N<sub>nuc</sub> underwent more diffusion during the precipitation due to their smaller size thus an increase in particles was observed. Precipitation from 1 to 3 mm can reduce the total particle concentration from  $2.2 \times 10^4$  to  $1.1 \times 10^4$  cm<sup>-3</sup>. The precipitation process not only influenced the concentration but also played a role in the size distribution of the particles. The meteorology parameters, such as wind,

helped in particle dispersion, and relative humidity triggered their secondary transformation. Air recirculation in the street canyon due to vehicular flow turbulence was observed in the study region.

The study investigated the role of emission sources in determining the concentration of nanoparticles in the urban roadside environment establishing relationships with seasonal factors, meteorology (wind parameters, precipitation and relative humidity), and gaseous pollutants, which helps to understand the dynamics of these particles in the atmosphere. The outcomes of the study can be used to estimate the health hazards these pollutants pose in the urban environment, especially those residing/ working near the road microenvironment.

### 5.5 Exposure estimation of nanoparticles in the study area

An analysis of quantification of particles related to health effects is also performed. The inhalable particle number (IPN) concentration over the study region varied from 10 million to 1 billion during different seasons for different breathing rates. The estimated IPN at a constant breathing rate of 12 liters per minute (LPM) showed that the concentration varies from about half a billion to 1 billion throughout the year. The nanoparticle deposition in the human respiratory tract estimated using. The deposition in the evening peak hours (EPH) during the winter season was >5 million particles. The annual average deposition of particles in the roadside environment (present study, 500 µg/year) was about 30% higher (330 µg/year) than a non-roadside ambient environment that is away from the roadside. The deposition of nanoparticles ranges between 0.29 and  $1.31 \times 10^{-3}$  µg/hr in winter, followed by spring (0.25-1.13× 10<sup>-3</sup>  $\mu$ g/hr), summer (0.01 × 10-3 - 1.13 × 10-4  $\mu$ g/hr), monsoon (0.02-1.1 × 10<sup>-3</sup>  $\mu$ g/hr) and autumn (0.01-7.88  $\times$  10<sup>-4</sup> µg/hr). The analysis reveals that an adult working 8 hours a day near the road environment in the study region experienced a nanoparticle deposition of 338  $\mu$ g/year. Further, on a 24-h scale (day) the residents residing near the roadside experienced a 3-times higher deposition (>1000 µg/year of nanoparticles) in the study region clearly indicating a linearity between the deposition of nanoparticles and the hours of exposure. The daily deposited mass concentration ranged between 0.02 and 0.04  $\mu$ g in the trachea region, followed by bronchiole < 0.05  $\mu$ g; in alveolar regions, the amount of particle mass deposited varied between 0.05 and 0.10  $\mu$ g/day. The rate of deposition in the alveolar region was higher for the age group of 8 to 21 years due to the increased breathing frequency. In terms of deposition rate in different regions of the human respiratory tract, 90% of particles get deposited in the alveolar regions, 6-8% in the bronchiole region, and 2% in the trachea region, highlighting that the deposition of nanoparticles was dominant in the alveolar regions. The deposition fraction of particles ranged between 0.02 to 0.04 in the alveolar region which was 4-5 times higher than in the bronchiole region. Compared to children and infants, adults experienced 30 to 40% higher nanoparticle deposition in the respiratory system. It may be noted that the model estimates of nanoparticle deposition may vary in real time on a day to day basis as the deposition depends on lung parameters and humidity. The new quantitative insights gained on the seasonal variation in the deposition of nanoparticles in humans residing near roadside conditions are crucial to estimate the human health risk potential, especially in winter when the nanoparticles and their deposition rates are significantly higher. These findings, with implications to improved air quality and public health are vital for formulating mitigation measures on a seasonal scale for exposure reduction to different age groups, hitherto unavailable, can lead to a better and sustainable future. The deposition of particles in different regions of the human respiratory system depends on the sizes of particles inhaled. The UFP deposition is found to be higher in the lower respiratory regions. The results on inhalable particle number concentration during different activities provide insights into the health effects on vulnerable populations in road environments during different seasons. These quantitative results obtained on a seasonal basis over an urban area in a megacity are crucial and valuable inputs to analyze the deposition of pollutants for an individual working/living in the vicinity of the road, and further to develop strategies for air quality, policy formulation (such as restrictions), and measures for combating air quality and climate change.

#### 5.6 Importance of the study

In the urban regions majority of the nanoparticles emissions come from the transportation sectors. The major transportation sector is road transport. The roadside microenvironment is the regions where the transportation emissions are found in higher concentration compared to the ambient environment. This study measured the concentration in the roadside microenvironment for year includes all the seasons and emission conditions is one of the first study of its kind in the study region. The study region is one of the highly urbanised city in the south Asia. The outcomes of the study help to represent further such cities similar to the study area. The results provide a detailed study during peak hours, different emission scenarios, day time and night time concentrations, role of gaseous precursors and role of meteorological parameters in determining the nanoparticles concentration. The study also examines the exposure of the nanoparticle's particles and their deposition. This study is one of the comprehensive study about the different aspects and features of the nanoparticles in the urban roadside microenvironments.

- The outcomes will enhance the knowledge about the nanoparticles in the urban atmosphere which is one of the major areas to focus for attaining clean air goals
- The study outcome prove the role of engine exhaust emission in the urban regions
- The study provides information about the different size engine exhaust emission which can be further used for the reduction of the emissions at the source.
- The results of the study help to estimate the health impacts of the residents living near the vicinity of the road as well as the people working in the vicinity of the road such as police personals, street vendors, delivery partners, drivers.
- The secondary transformation of the particle can also be studied using the primary emissions.
- The insights of the study help in policy formulation for the number concentration as suggested by the WHO for detecting and reducing the nanoparticles emissions.

• The results of the study help the policy makers to develop the mitigation measures for source specific pollution which provides more efficiency of the control measures.

These are some of the major outcomes of the study which emphasise the importance of the study.

## 5.7 Scope of further study

The study on the atmospheric nanoparticles is one of the grey areas where majority of the information's are found missing due to limited studies paves way for more further studies in this direction.

- The fuel-based nanoparticles emission in the real-world driving conditions can be studied for different vehicles to estimate the nanoparticles concentration
- Different vehicles (Two wheelers, four-wheelers and heavy vehicles) based studies can be performed in future to know the emissions scenario.
- More further studies can be conducted during different hours of the day such as day and night time, peak and non-peak hours and weekdays and weekends.
- The other microenvironments in urban regions such as tunnels, airport zones, industrial areas are the major other environments which needs to be explored.
- The Indoor environments is another major area where the concentration of nanoparticles play a crucial role in human health.
- The source apportionment of atmospheric nanoparticles is another grey area which needs to be explored for identifying the different complex sources and their composition in the atmospheric nanoparticle concentration.
- Apart from model studies clinical epidemiological studies should be conducted to assess the exact particle deposition occurred in the respiratory system.

## 5.8 Summary

The discussions in this chapter concludes the different results obtained from the study which is designed based on the objectives. The sections also cover the importance of the study and also the future scopes of the study which can be conducted in the further studies.

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## ANNEXURES

Annexure Table.1 F-Test for variance obtained between different phases in two study periods.

Period- I			Period-II		
Phase	BR	DR	Phase	BD	AD
Mean	32121	24372	Mean	24192	42461
Variance	223711144	146426271	Variance	114995765	373545594
Observations	409	918	Observations	409	192
Df	408	917	Df	408	191
F	2		F	0	
P(F<=f) one-			P(F<=f) one-		
tail	0		tail	0	
F Critical one-			F Critical		
tail	1		one-tail	1	
Phase	DR	AR	Phase	AD	DRII
Mean	24372	25265	Mean	42461	39532
Variance	146426271	202240814	Variance	373545594	389046175
Observations	918	596	Observations	192	432
Df	917	595	Df	191	431
F	1		F	1	
P(F<=f) one-			P(F<=f) one-		
tail	0		tail	0	
F Critical one-			F Critical		
tail	1		one-tail	1	
Phase	BR	AR	Phase	BD	DRII
Mean	32121	25265	Mean	24192	39532
Variance	223711144	202240814	Variance	114995765	389046175
Observations	409	596	Observations	409	432
Df	408	595	Df	408	431
F	1		F	0	
P(F<=f) one-			P(F<=f) one-		
tail	0		tail	0	
F Critical one-			F Critical		
tail	1		one-tail	1	

Phase	Nnuc	Nsatk	Nlatk	Nacc	Ntotal
BR	3089	1987	1771	658	60451
DR	1970	1488	1736	1055	47580
AR	2286	1882	1579	471	49283
BD	1557	1759	2121	972	52114
AD	1708	2093	3356	1722	77391
DRII	1445	2344	3568	1625	78632

Annexure Table. 2 95% confidence interval values in different size ranges and phases

Annexure Table. 3 Levene test for variance and significance in different size ranges and phases.

Period – I			Period-II			
N <sub>nuc</sub>	F value	Pr(>F)	N <sub>nuc</sub>	F value	Pr(>F)	
2	42.02361	1.37E-18	2	2.812837	0.060392	
1920	NA	NA	1330	NA	NA	
Nsatk	F value	Pr(>F)	N <sub>satk</sub>	F value	Pr(>F)	
2	15.02892	3.34E-07	2	14.48712	5.97E-07	
1920	NA	NA	1330	NA	NA	
N <sub>latk</sub>	F value	Pr(>F)	N <sub>latk</sub>	F value	Pr(>F)	
2	4.673534	0.009446	2	58.66576	3.83E-25	
1920	NA	NA	1330	NA	NA	
Nacc	F value	Pr(>F)	Nacc	F value	Pr(>F)	
2	43.02785	5.25E-19	2	37.26233	1.80E-16	
1920	NA	NA	1330	NA	NA	
N <sub>total</sub>	F value	Pr(>F)	N <sub>total</sub>	F value	Pr(>F)	
2	8.631657	0.000185	2	35.33078	1.12E-15	
1920	NA	NA	1330	NA	NA	
AT	F value	Pr(>F)	AT	F value	Pr(>F)	
2	55.5216	3.62E-24	2	3.688203	0.025275	
1920	NA	NA	1321	NA	NA	
RH	F value	Pr(>F)	RH	F value	Pr(>F)	
2	21.97782	3.65E-10	2	2.019814	0.13309	
1920	NA	NA	1321	NA	NA	
SR	F value	Pr(>F)	SR	F value	Pr(>F)	
2	6.292932	0.001888	2	8.762333	0.000166	
1920	NA	NA	1321	NA	NA	

## LIST OF PUBLICATION AND THEIR PROOFS

Urban Climate 51 (2023) 101625





## Roadside measurements of nanoparticles and their dynamics in relation to traffic sources in Delhi: Impact of restrictions and pollution events

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ARTICLE INFO

Keywords: Nano pollutants Number concentration Traffic emission Ultra fine particles Urban aerosols

### ABSTRACT

Due to rapid urbanization, Delhi experiences frequent pollution events, and the particulate matter load exceeds the prescribed limit often. This study analyzes nanoparticles (10 to 1090 nm) during different emission scenarios, seasonal and meteorological conditions in two phases: April to June 202] (Period I) and October to November 202] (Period II). Period I experienced around 31% less concentration of particles ( $\sim$ 2.4 × 10<sup>4</sup> cm<sup>-2</sup>) due to lockdown restrictions and, on the other hand, particle concentration increased by 35% compared to normal conditions due to the sudden rise in firework emissions in Period II. Except for the post-Diwali phase (10<sup>4</sup> cm<sup>-3</sup> to 10<sup>5</sup> cm<sup>-3</sup>), the concentrations lie between 10<sup>3</sup> cm<sup>-3</sup> and 10<sup>5</sup> cm<sup>-3</sup>. The Aitken modes contribute 10 to 30% of total concentration in both periods. Particles in nucleation and accumulation modes contribute 30 to 40%, 20 to 30%, 15 to 25%, and 35 to 50% in Periods I and II, respectively. Number concentration-based studies are essential for estimating the potential impacts on human health due to air pollution. The study provides information regarding vehicle emission-based particle concentration under various emission scenarios in urban cities, which is crucial for estimation of emissions, health impact assessment, future policy formulation and strategy measures.

### 1. Introduction

In an urban environment, traffic-related air pollution occurs due to a mixture of both particulate matter and gases in a complex

Abbreviations: Acc, Accumulation mode; AD, After Diwali; AR, After Restriction; BD, Before Diwali; BR, Before Restriction; cm-3, per Cubic Centimeter; CNG, Compressed Natural Gas; COVID-19, Coronavirus Disease - 2019; CPC, Condensed Particle Counter; DMA, Differential Mobility Analyzer; Dp, Diameter of the Particle; DR, During Restriction; DR-II, During Restriction II; DTU, Delhi Technological University; E, East; E-Rickshaws, Electric rickshaws; Eqn, Equation; GMD, Geometric Mean Diameter; HCV, Heavy Commercial Vehicle; hpa, hecto pascal pressure unit; Hrs, Hour; ISO, International Organization for Standardization; Latk, Large aitken mode; LCV, Light Commercial Vehicle; LDMA, Large Differential Mobility Analyzer; LPM, Liters Per Minute; m, Meter; MBq, Megabecquerel; m/s, Meter per Second; N, North; NA, Not Applicable; NCT, National Capital Territory; nm, nanometer; Nuc, Nucleation mode; PNC, Particle Number Concentration; RH, Relative Humidity; Satk, Small Aitken Mode; SD, Standard Deviation; S. No, Serial Number; SR, Solar Radiation; SVOC, Soluble Volatile Organic Compounds; AT/Temp, Ambient Temperature; Total, Total Particle Number Concentration; UFP, Ultrafine Particles; USA, United States of America; VOC, Volatile Organic Compounds; W/m-2 Watt per square meter; WS, Wind Speed.

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## Size resolved particle contribution to vehicle induced ultrafine particle number concentration in a metropolitan curbside region



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### HIGHLIGHTS

Seasonal nanoparticle number concentration (PNC) in Delhi varies from 10<sup>4</sup> to 10<sup>6</sup> cm<sup>-2</sup>.

Particles in nucleation mode contribute >30% to PNC in summer, spring and monsoon.

Accumulation mode particles contribute >35% to PNC in winter and autumn.

Higher relative humidity favors coagulation resulting in higher concentration of Nacconstruction.

Correlation between N<sub>acc</sub> and Carbon Monoxide is higher due to transport emissions.

### ARTICLE INFO

Keywords: Particle number concentration Seasonal variation Transport emissions Urban pollution and meteorology Implications

### ABSTRACT

The concentrations and behavior of nano particles (10–1000 nm) in Delhi, a densely populated megacity, in different seasons (winter, spring, summer, monscon, and autumn) are examined, for the first time. The concentration of particles is classified into four different sizes as N<sub>mac</sub> (10–30 nm, nucleation), N<sub>mac</sub> (30–50 nm, small Altken), N<sub>mac</sub> (50–100 nm, laspe Altken), and N<sub>mac</sub> (100–1000 nm, accumulation mode), and the total (10–1000 nm) particle number concentration (PNC) as N<sub>totab</sub>. PNC ranges between 10<sup>4</sup> cm<sup>-2</sup> and 10<sup>6</sup> cm<sup>-2</sup> over Delhi during the year, and the highest concentration occurs in winter. Winter concentration is 2, 1.6 and 1.3 times higher than monsoon, summer, autumn and spring concentrations, respectively. N<sub>trac</sub>, N<sub>mac</sub>, N<sub>bak</sub> and N<sub>acc</sub> and their respective contributions to total PNC exhibit significant seasonal variations. Uniter N<sub>tota</sub>, and N<sub>acc</sub> contribute more to total due to coagulation, with N<sub>hec</sub> alone contributing >40% to total PNC. N<sub>trac</sub>, N<sub>mac</sub>, N<sub>mac</sub>, N<sub>mac</sub>, N<sub>mac</sub>, N<sub>mac</sub>, and N<sub>hack</sub> direct primary emissions from engine exhans produce a prominent double hump structure during moning and evening peak hours in winter and autumn. PNC and their contributions exhibit day-night variations as they are influenced by variations in emission sources and meteorological parameters (wind speed, relative humidity, temperature, solar radiation and boundary layer height) between day and night. Carbon monoxide correlates positively with N<sub>hacc</sub> in all seasons (R<sup>2</sup> ≥ 0.5) as fossil fuel emission is a predominant source for gases and particles in the study environment. These quantitative results on seasonal variations of nano particles and air pollutants together with the knowledge on seasonal variations in meteorological parameters and atmospheric dynamics provide a foundation which can positively contribute better to the urban planning and devising mitigation measures aimed at improving air quality and public health.

### 1. Introduction

The air quality of any region depends on the quantity and quality of different particles present in the atmosphere (Yadav et al., 2021; Zhao et al., 2020). These particles in the atmosphere can cause several health impacts, including cardiovascular and pulmonary diseases (Kanawade et al., 2020; Kaur et al., 2005; Sharma et al., 2024; Schraufnagel, 2020). The particles present in the air vary based on size, shape, and composition, and their potential to create health impact also depends on these characteristics (Gao et al., 2015; Gurjar et al., 2010; Kumar et al., 2011;

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Chemosphere 366 (2024) 143470

Seasonal variation of particle number concentration in a busy urban street with exposure assessment and deposition in human respiratory tract

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### HIGHLIGHTS

### Ultrafine particles (UFPs) contribute around 60-80 % to total particle number concentration (PNC) over Delhi.

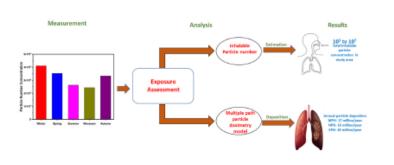
- Inhalation particle number concentration varies between 0.5 and 1 billion over Delhi during the year.
- The particle deposition ranges from 0.43 to 0.26 μg/min over Delhi during the year.
- The concentration of particles in the urban roadside is nearly 30% higher than an environment away from the road.
- The results are crucial for initiating mitigation measures aimed to improve air quality and public health.

### ARTICLE INFO

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Keywords: Air pollution exposure Human health Particle dosimetry model Peak vs. non-peak hours Ultrafine particles

## GRAPHICAL ABSTRACT



### ABSTRACT

Ultrafine particles (UFP) associated with air quality and health impacts are a major concern in growing urban regions. Concentrations of UFP (particles of size between 10 and 100 nm) and accumulation mode (N<sub>acc</sub>) (particles of size >100 and up to 1000 nm), are analyzed over a highly polluted megacity, Delhi, in conjunction with vehicular flow density, during peak (morning, and evening) and non-peak hours. UFP contributes >60% to total particle concentration during autumn and monsoon. UFP concentrations are about 50,000 particles per cm<sup>3</sup> in winter which reduces to about 25,000 particles during monsoon. N<sub>acc</sub> are about 20,000 (winter) and 10,000 (monsoon) particles per cm<sup>3</sup>. UFP concentration and N<sub>acc</sub> during peak hours are at least twice higher than those obtained in non-peak hours, confirming the dominant influence of emissions from vehicular exhaut in the study region. Seasonal analysis of UFP size distribution reveals that direct emissions dominate the particle concentrations during winter and autumn, whereas new particle formation and particle deposition in the human respiratory tract using Multiple Path Particle Dosimetry (MPPD) model, performed for the first time, shows that the order in which these particles deposit in the human respiratory tract using Multiple Path Particle Dosinter (MPPD) model, performed for the first time, shows that the estimated inhalable particles concentration (IPN) varies between 0.5 and 1 billion. Results on the IPN during activities classified from light (walking), medium, heavy, very heavy to severe (long-distance running) provide insights into health effects on vulnerable populations. These quantitative results obtained over a megacity on

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## **ARTICLE IN PRESS**

Atmospheric Pollution Research xxx (xxxx) xxx



## Influence of local meteorology and gaseous pollutant emissions on atmospheric nanoparticle concentrations in a pedestrian way in urban region

ABSTRACT

### Kanagaraj Rajagopal<sup>a</sup>, S. Ramachandran<sup>b</sup>, Rajeev Kumar Mishra<sup>a,\*</sup>

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### ARTICLE INFO

#### Keywords: Air pollution Engine exhausts Gaseous emissions Meteorology Ultrafine particles Urban climate

### The roadside environment is one of the major sources of nanoparticle emission in the urban regions. The complex mixture of different pollutants makes it hard to understand the dynamics and behavior of nanoparticles. The study aims to analyze the dynamics of atmospheric nanoparticles ranging from 10 to 1000 nm alo ing with local meteorological conditions and gaseous pollutants in a pedestrian way at a busy street in Delhi, a megalopolis in India, during all five major seasons of the study area. Nucleation mode particles were higher during the spring on, whereas the contribution of Aitken mode particles dominated the total particle number concentra u in the rest of the seasons. Due to the influence of vehicular sources, rush hour concentrations were higher than during non-rush hours. Thus, the diurnal pattern in particle concentration strongly coincided with emissions ciated with the vehicular flow. During the winter season, the average total particle number concentrati was observed to be maximum (4.1 $\times$ 10<sup>4</sup> cm<sup>-3</sup>) with a higher surface area of particles of 3.5 $\times$ 10<sup>-3</sup> mm<sup>2</sup> m<sup>-3</sup>. Compared to the monsoon season, the concentration of $NO_X$ was 5 times higher in winter. The boundary layer height in the study region ranged from 600 to 2400 m during different seasons, and the maximum ventilation coefficient was observed to be $> 3000 \text{ m}^2 \text{s}^{-1}$ during summer. Precipitation reduced the concentration of particles by half, from 2.2 x 10<sup>4</sup> to 1.1 x 10<sup>4</sup> cm<sup>-3</sup>, due to wet scavenging. The study revealed that the concentrations of particles depend not only on primary emissions but also are influenced by local meteorology and other co-emitted pollutants. Understanding the dynamics of atmospheric nanoparticles in urban roadside environments as outlined in this study is crucial to devise necessary mitigation measures for people residing near the road in order to reduce health impact and improve air quality.

### 1. Introduction

Rapid urbanization and economic growth lead to serious air pollution issues in the Indian subcontinent. The complex mixture of sources contributes to poor air quality in urban cities, with emissions from traffic-related sources dominant among them (Sharma et al., 2024). Exhaust and non-exhaust (break, tyre wear and dust resuspension) emissions are two major emissions classifications from the transportation sector. Traffic-related exhaust emissions originate from different categories of vehicles, such as two, three-wheelers, person al/commercial cars, buses, and heavy vehicles (trucks, lorries, and crames) (Belkacem et al., 2020; Pandey et al., 2022; Rogula-Kozlowska, 2014; Wu et al., 2021). Engine exhaust emits a wide range of pollutants, such as particulates and gaseous pollutants, based on the type of ignition, fuel used etc. (Myung and Park, 2012; Rajagopal et al., 2023; Yu et al., 2022). Emission norms for vehicles in India are based on the Bharath Stage VI standard with an emission limit for particulate matter (PM) < 4.5 mg/km and nitrogen oxide (NO<sub>X</sub>) < 60 mg for petrol and < 80 mg for diesel engines (Rao et al., 2021). Apart from these emissions, the engine exhaust also emits particles that range from a few nanometers to micrometers (Patton et al., 2017; Sabaliauskas et al., 2012; Zhu et al., 2021). The nanoparticle emissions from the engine exhaust are broadly, 2021). The nanoparticle of (N<sub>nuc</sub>, 1–30 nm), Aitken mode (N<sub>atko</sub>, 30–100 nm), and accumulation mode (N<sub>acco</sub>, 100–1000 nm) (Kumar

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# Are Delhi residents exposed to lesser particle number concentration due to the firework ban in the city?

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### Abstract

Diwali, the fireworks festival of India, adds more atmospheric particles within a short period of time and deteriorates the air quality. Short-term policies like banning crackers during fireworks festivals can help improve urban air quality. The present study analyzed particle number concentration, ranging from 10 to 1000 nm, in 2021 and 2022. A reduction in the concentration of particle number concentration (from  $3.8 \times 10^4$  cm<sup>-3</sup> to  $3.1 \times 10^4$  cm<sup>-3</sup>) was observed due to the ban on crackers in the urban city of Delhi. The concentration range changes from  $10^5$  cm<sup>-3</sup> to  $10^4$  cm<sup>-3</sup>. The contribution of different size ranges, Nucleation (10 to 30 nm), Aitken (30 to 100 nm), and Accumulation (100 to 1000 nm) are analyzed. During Diwali day, the Accumulation mode particle concentration on Diwali (During ban on firecrackers) was reduced by about 18%, i.e., 1.6 million particles per day. The study results show that emissions in urban regions can be reduced significantly by proper implementation of policy and participation from citizens. Reducing particle emissions paves the way for air quality improvement, health impact mitigation, and sustainability. Sustainability goals focus on clean air for all, and health improvement in polluted regions as interim goals, that can be achieved by implementing proper mitigation measures, which consequently help fight climate change.

Keywords Diwali emission · Delhi air quality · Firework emissions · Particle number concentration

### Introduction

Diwali is one among the major festivals in India, and it is celebrated by bursting of crackers and lighting of lamps throughout the country every year (Ganguly 2015; Garg and Gupta 2018; Yadav et al. 2019, 2022a). The firecrackers used during the festival use different chemicals such as aluminum, sulfur, potassium nitrate, barium nitrate,

### Highlights

 Firework emissions contribute up to 83% of accumulation mode particles in total particle number concentration.

3. Crackers ban reduced inhalable particle concentration on Diwali day, leading to less exposure.

 The percentage contribution of different size ranges to total particle number concentration is based on the sources present.

Rajeev Kumar Mishra rajeevkumarmishra@dtu.ac.in charcoal, manganese, strontium nitrate, potassium, and iron dust powder as a composition material for the manufacturing of firecrackers (Nishanth et al. 2012; Perrino et al. 2011; Sateesh et al. 2018). The bursting of crackers emits different types of atmospheric pollutants, such as particulate matter, ultrafine particles, gaseous pollutants, and toxic metals of significant quantity (Chatterjee et al. 2013; Izhar et al. 2018; Vaghmaria et al. 2018; Yerramsetti et al. 2013). These particles stay in the atmosphere for a few days, causing the formation of toxic smog (Cetin 2015; Ertugrul et al. 2019; Ozenen Kavlak et al. 2021). The pollutants are hazardous and cause serious health effects (Kanawade et al. 2014).

Delhi is one of the highly urbanized cities which experiences frequent pollution events. Air quality in the city remains in the poor category for most of the year (Agarwal et al. 2020; Mishra et al. 2016; Mohan and Kumar 2022). The air quality consistently exceeds the national standards (Garg and Gupta 2018; Goyal et al. 2021; Kanawade et al. 2020). Diwali festival usually falls under October and November, the start of the winter season in Delhi (Garg and Gupta 2018; Ghei and Sane 2018). The winter period

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A firework ban can reduce up to 20% to 22% of Aitken and Accumulation mode particles on the Diwali festival.

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# **Curriculum Vitae/Brief Profile**

# Kanagaraj Rajagopal

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To work with an organization which gives me ample opportunity to learn and grow along with the organization and to prove myself worthy of shouldering the responsibilities assigned to me

## EDUCATIONAL DETAILS

Education	Institute	Percentage	Year of completion
		Obtained	
Ph.D.	Delhi		
Environmental	Technological	_	2024(expected)
Engineering	University		
M.Tech Green Energy	Pondicherry	8.2 (CGPA)	2018
Technology	University		
<b>B.E</b> Environmental	Park College of	6.9(CGPA)	2015
Engineering	Technology		
HSC	RKR Grks MHSS	78.41	2011
SSLC	RKR Grks MHSS	82.8	2009

## WORK EXPERIENCE

YEAR	Role and Organization
May 2015 to June 2016	Jr. Environmental Engineer
2018 (Nov)- 2019(July)	Project Associate/JRF – IIT-Roorkee
2019 (Sep) – 2020 (Jan)	JRF – Delhi Technological University

## AREA OF INTEREST

- Ultrafine particles.
- Urban aerosols.
- Gridded emission inventory.
- Air & Noise pollution monitoring.
- Environmental impact assessment.
- Carbon emissions, climate change.

Software Skill

- Arc GIS / QGIS
- Origin
- MATLAB

- R statistical tool
- AirQ+
- Multiple path particle dosimetry model.

## **PROJECTS / DISSERTATION**

- B.E Dissertation EIA on Residential Building Construction in Coimbatore city.
- Diwali Air and noise pollution study in Coimbatore with Tamilnadu Pollution Control Board (TNPCB) in 2013 & 2014.
- Coimbatore city traffic noise pollution study in 2013&2014.
- M.Tech Dissertation Photocatalytic degradation of Textile dyes using solar concentrating helical tubular reactor.
- Ph.D. Study of nano-size particle dynamics in urban road microenvironment in Delhi.

## **RESEARCH PROJECTS**

- 1. Megacity Delhi atmospheric emission quantification, assessment and impacts (DelhiFlux) URL (<u>https://www.urbanair-india.org/delhiflux</u>) funded by Ministry of Earth Sciences.
- 2. The air pollution study during the Odd-Even Scheme in Delhi was funded by DTU.
- 3. Proxy Relationship of Ultrafine Particles Number Concentration, New Particle Formation and Its Growth Rate in Transport Microenvironment in Delhi funded by CPCB.

## SEMINAR / TRAINING/WORKSHOP

- Participated in one day workshop at Park College of Technology, "Carbon sequestration and Climate Change"
- Participated in a two-day seminar on "Emerging trends in the environment" at Kochi conducted by Kerala Environmental Congress.
- Participated in the Smart City conference Organized by Park College of Technology in the year 2015.
- Participated in the MODEL presentation on "GREEN BUILDING DESIGN" at TNAU and bagged 2nd prize (2013).
- Participated in IDEAL IDEA contest conducted by TNAU and WON IInd prize (2014).
- Attended In-plant training in ABC Techno Labs pvt ltd, Coimbatore
- Participated in the MODEL presentation contest conducted by TRP Engineering College, Trichy.
- Participated in the International Workshop on Energy Materials and Devices (IWEMD 2018) organized by Pondicherry University.
- Participated in an International Workshop on Sustainable Building Materials and Construction Conducted by Pondicherry University.
- Participated in Biogas plant Installation and operation workshop at CGET.
- Participated in TNAU Eco-fest 2017 and won First Price in Paper presentation also, the team bagged the Overall Champion trophy for Pondicherry University.

## INTERNATIONAL CONFERENCE

**1.Kanagaraj Rajagopal**, Kanak Sharma, Mayank Goyal, T. Vijayakumar, Rajeev Kumar Mishra (2023). Prediction of fuel usage policy implementation with household air pollution and contributions towards sustainability in Indian subcontinent. SERB Sponsored International Conference on "Technological Advancements in Materials and Manufacturing for Industrial Environment from 5<sup>th</sup> to 6<sup>th</sup> may at KPRIET.

2. Kanagaraj Rajagopal, S. Ramachandran, Rajeev Kumar Mishra (2022). Measurements of Size Resolved Nanoparticle Concentration and its Distribution in Polluted Urban Environment during Induced Firework Event.7<sup>th</sup> Indian International Conference on Air Quality Management (IICAQM 2022), organized in association with IIT Madras, IIT Kharagpur, IIT Guwahati, University of California (USA), University of Bath (UK), Asian Institute of Technology (Thailand), Mahidol University (Thailand) and Australian National University (Australia) from 27 November to 1<sup>st</sup> December 2022 at IIT Madras, India.

**3. Kanagaraj Rajagopal**, S. Ramachandran, Rajeev Kumar Mishra (2023). Traffic-Induced ultrafine particle concentration in megacity Delhi. 8<sup>th</sup> Indian International Conference on Air Quality Management (IICAQM 2023) jointly organized by IIT Madras, IISc Bangalore, IIT Guwahati, Australian National University (Australia), Mahidol University (Thailand), University of Bath (UK), Asian Institute of Technology (Thailand), University of California, Riverside (USA), IIT Bombay and IIT Kanpur from 4<sup>th</sup> to 8<sup>th</sup> December 2022 at IISc Bangalore, India.

**4.** Shailendra Kumar Yadav, Monika Sharma, **Kanagaraj Rajagopal**, Vignesh Mohan, Veerendra Sahu, Rajeev Kumar Mishra, Bhola Ram Gurjar (2023). Investigating the effects of Delhi's odd-even vehicle registration policy on particle number concentration. International conference on waste recycling and environmental technology (WRET-20254) from 8<sup>th</sup> to 9<sup>th</sup> February at BBAU.

### MEMBERSHIP

- **1.** Life member of the Air Quality Management Association.
- 2. Member of Green Energy Forum.
- **3.** Member of the Society for Indoor Environment.
- **4.** Member of the European Geoscience Union.
- 5. Member of Environmental Engineers Association.
- **6.** Member of The Aerosol Society.

## BOOK CHAPTERS PUBLISHED

- Shailendra Kumar Yadav, Rajagopal, K., Priya, A.K., Sharma, G.D., 2020. Smart Waste Management and Energy Extraction from Waste in Indian Smart Cities – A Review, in: Contaminants and Clean Technologies. CRC Press, pp. 321–330. https://doi.org/10.1201/9780429275852-19.
- Shailendra Kumar Yadav, Rajagopal, K., 2019. Hydroponic Treatment System Plant for Canteen Wastewater Treatment in Park College of Technology, in: Zero Waste. CRC Press, pp. 187–202. <u>https://doi.org/10.1201/9780429059247-12</u>.

 Kumar, P, Rajagopal, K., 2019. Plastic for Sustainability- Fundamentals of Plastic Waste Management. DBH publishers and distributors, volume-12, pages 142-144 <u>https://doi.org/10.13140/RG.2.2.32476.21128</u>.

## **RESEARCH PAPER PUBLISHED**

- I. Rajagopal, K., Ramachandran, S., and Mishra, R.K (2024). Influence of local meteorology and gaseous emission on atmospheric nanoparticle concentrations in the pedestrian way in the urban region. *Atmospheric pollution research*. 102358. <u>https://doi.org/10.1016/j.apr.2024.102358</u>.
- 2. Rajagopal, K., Ramachandran, S., & Mishra, R.K (2024). Seasonal variation of particle number concentration measurements in an urban busy street exposure assessment and deposition in human respiratory tract. Chemosphere, 366,143470 <u>https://doi.org/10.1016/j.chemosphere.2024.143470</u>.
- 3. Rajagopal, K., Ramachandran, S., & Mishra, R.K. (2024). Size resolved particle contribution to vehicle induced ultrafine particle number concentration in a metropolitan curbside region. Atmospheric Environment, 337, 120773. https://doi.org/10.1016/j.atmosenv.2024.120773.
- 4. Rajagopal, K., Mohan, V. & Mishra, R.K. (2024). Are Delhi residents exposed to lesser particle number concentration due to the firework ban in the city? Air Quality, Atmosphere & Health, 1-11 <u>https://doi.org/10.1007/s11869-024-01532-3</u>.
- 5. Rajagopal, K., Ramachandran, S., & Mishra, R.K. (2023). Roadside measurements of nanoparticles and their dynamics in relation to traffic sources in Delhi: Impact of restrictions and pollution events. Urban Climate, 51, 101625. <u>https://doi.org/10.1016/j.uclim.2023.101625</u>.

## TECHNICAL MAGAZINES PUBLISHED

- Operation and Maintenance of STP- an Engineering view (Water Today magazine).
- Operation and Maintenance of Sewage Treatment Plant Non-Engineering View (Water Today magazine).
- Waste Water Generation Estimation and Sustainable Sanitation Development (Water Today magazine).

## EDITORIAL IN NEWSPAPER

**Telegraph UK:** Nanoparticle emissions from Delhi's transport sector could trigger significant health risks: Study – Business Telegraph

**Gulf News:** Nanoparticle emissions from vehicles in Delhi may raise health risks: Study | India – Gulf News

PTI: Delhi transport sector could trigger significant health risk.

PTI Bhasha: दिल्ली मेंसड़कों पर वाहनों सेननैीकण का उत्सर्जनर्ज स्वास्थ्य जोखिमपदा कर सकता है: अध्ययन IANS: Nanoparticle emissions from vehicles in Delhi may raise health risks: Study

**Money control:** Delhi sees rise in nanoparticle emissions, researchers warn of serious health risks **Indian Express:** 'Nanoparticles from vehicle fumes can cause acute illness': Study looks at 2021 Delhi data over two periods

Times of India (Delhi): 'After lockdown, 35% rise in nanoparticle emissions' | Delhi News -

**Times of India Times of India (Nagpur):** Nanoparticles pose a mega risk to Delhi | Nagpur News - Times of India

Mid-day: Nanoparticle emissions from vehicles in Delhi may raise health risks: Study

**Economic Times:** Nanoparticle emissions from Delhi's transport sector could trigger significant health risks: Study - The Economic Times

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Hindustan: दिल्ली की हवा मेंननैोकण का खतरनाक स्तर, वज्ञैानि कों नेकि या आगाह, कि न्हेंज्यादा जोखि म?

**News9:** Delhi witnesses rise in nanoparticle emissions from transport sector that may lead to serious health risks: Study

The Print (Hindi): दिल्ली मेंसड़कों पर वाहनों सेननैोकण का उत्सर्जनर्ज स्वास्थ्य जोखिमपदा कर सकता है: अध्ययन – ThePrint Hindi

NewsClick (Hindi): दिल्ली मेंसड़कों पर वाहनों सेननैोकण का उत्सर्जनर्ज स्वास्थ्य जोखिमपदैा कर सकता है: अध्ययन। न्यज़ू क्लि क

Business Standard: Nanoparticle emissions from vehicles in Delhi pose health risks: Study

Dainik Bhaskar: Nanoparticle emissions from vehicles in Delhi may raise health risks: Study – Bhaskar Live

Navbharat Times: Delhi Air Pollution, दिल्ली की हवा मेंतरै रहा सांस का 'दश्मन', जानेंक्यों डरा रही येनई स्टडी

**The Patriot:** Study Reveals Potential Health Risks Due to Nanoparticle Emissions from Delhi's Transportation Sector - The Patriot

**Climate Samurai:** Delhi Grapples with Soaring Levels of Hazardous Nanoparticles in Urban Air: study – Climate Samurai

Hindi Nav Samachar: दिल्ली की शहरी हवा में ननैोकणों के बढ़तेस्तर की चेतावनी – नवसचंर समाचार .कॉम Latestly: देश की खबरें| दिल्ली मेंसड़कों पर वाहनों सेननैोकण का उत्सर्जनर्ज स्वास्थ्य जोखि मपदैा कर सकता है: अध्ययन | LatestLY हिन्दी

Dynamite News: दिल्ली मेंसड़कों पर वाहनों सेननैोकण का उत्सर्जनर्ज स्वास्थ्य जोखि म पदाै कर सकता है: अध्ययन - डाइनामाइट न्यज़्

Siasat Daily: Nanoparticle emissions from vehicles in Delhi may raise health risks: Study

Lokmat Times: Nanoparticle emissions from vehicles in Delhi may raise health risks: Study

Janta Se Rishta: दिल्ली मेंवाहनों सेनि कलनेवाले ननैोकणों सेस्वास्थ्य जोखि म बढ़ सकता है: अध्ययन | Nanoparticle emissions from vehicles in Delhi may raise health risks

IBC24: दिल्ली मेंसडकों पर वाहनों सेननै ोकण का उत्सर्जनर्ज स्वास्थ्य जोखि म पदैकर सकता है: अध्ययन |

ETV Bharat: Nanoparticle emissions from Delhi's transport sector could trigger significant health risks: Study

**Newsdrum:** Nanoparticle emissions from Delhi's transport sector could trigger significant health risks: Study

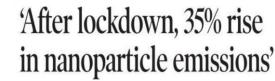
### DECLARATION

I hereby declare that all the information furnished by me is true to the best of my knowledge and belief.

R.Howen

[Kanagaraj R]

# Impact of the Research in online and print media (National and International)



### Kushagra.Dixit@timesgroup.com

New Delhi: Scientists have found that after the Covid lockdown eased in Delhi, the emission of nanoparticles, which are super-fine pollutants, increased by 35% on the city's roadside. Nanoparticles are fine pollutants

which are about 1/1000th of particulate matter. They cannot be filtered out by N-95 masks.

Scientists in the department of environmental engineering at Delhi Technological University (DTU) and the space and atmospheric sciences division of the Physical Research Laboratory in Ahmedabad did a study on the roadside nanoparticles in Delhi and their impact on health and air quality.

The study, called "Roadside Measure ments of Nanoparticles and their Dyna-mics in Relation to Traffic Sources in Delhi: Impact of Restrictions and Pollution Events", found that after the lockdown, the nanoparticles had increased.

The researchers led by Dr Rajeev Kumar Mishra, faculty in the department of environmental engineering in DTU, professor S Ramachandran of the Physi-



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cal Research Laboratory, and Kanagaraj Rajagopal, a rese-arch scholar in the advance air and acoustics research laboratory in DTU, identified the concentration of the nanoparticles present on the roadside. The study includes a series

of events in 2021, such as the lockdown, pollution events, Diwali and the time period before and after these events. It aims to identify the reduction of nano pollutants based on the emission source In Delhi, 10.000 to 10.00.000 nanoparti cles are found in a cubic centimetre of air. During the lockdown period, when vehicles on the road were reduced to around 50%, the nanoparticle concentration was reduced by 31 %. At the same time, events such as Diwali increased nanoparticle concentration up to 35 %, compared to usual emissions," Mishra said, adding that the nanoparticles in the range of 10 to 100-nm size were directly from vehicle engine exhaust that were released on the roadside.

The study said that when windspeed is high, the nano pollutants got dispersed to surrounding regions, increasing exposure among residents living near the road.

"For a city like Delhi, residential zones adjacent to the road will automatically mean higher exposure for residents. The nanoparticles are more vulnerable in terms of human health be cause they are much smaller than PM2.5 or PM10." the study pointed out.

These nanoparticles are 600 times smaller than the thickness of human hair and can penetrate deep into the lungs and blood stream.

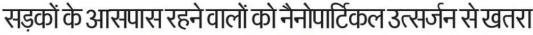
"These nanoparticles have the potential to penetrate our bloodstream and can be deposited in different parts of the human body, including the brain.... People working or living near the road, such as police personnel, street vendors, drivers, motorcyclists, delivery personnel and the urban poor living near the road are more exposed to them," Mishra said.

The study also suggests that policy formulation towards the concentration of the particles was required for reducing their emission from engine sources to mitigate the impact these particles have on the atmosphere and human health.



नई दिल्ली, एजेंसी। दिल्ली में सड़क किनारे वातावरण में नैनोकण का खतरनाक स्तर पाया गया है, जिसका सीधा संबंध वाहनों से निकलने वाले धएं से है। इससे स्वास्थ्य संबंधी चिंताएं बढ रही हैं। यह जानकारी एक अध्ययन के दौरान सामने आई। अध्ययन में बताया गया कि मनुष्य के बाल से 600 गुना बारीक होने के कारण ये कण फेफडों. रक्त में प्रवेश कर सकते हैं। पत्रिका 'अर्बन क्लाइमेट' में

प्रकाशित इस अध्ययन को उत्तर-पश्चिमी दिल्ली में बवाना रोड पर किया गया था, जो दिल्ली को रोहतक से जोडता है। दिल्ली प्रौद्योगिकी में विश्वविद्यालय पर्यावरण अभियांत्रिकी विभाग के सहायक प्रोफेसर राजीव कुमार मिश्रा ने कहा कि अध्ययन में पाया गया है कि जब हवा की गति तेज होती है तो ये कण सड़क के नजदीकी क्षेत्रों में फैल जाते हैं। इससे जोखिम बढ़ जाता है।



### शिप्रा सुमन॰ बाहरी दिल्ली

राजधानी में बढ़ते वायु प्रदूषण पर दिल्ली प्रौद्योगिकी विश्वविद्यालय ने सड़क किनारे के वातावरण पर शोध किया है। इसमें पता चला है कि सड़क किनारे के क्षेत्रों में रहने लोगों को नैनोपार्टिकल वाले उत्सर्जन से जोखिम बढ सकता है। डीटीयू के पर्यावरण इंजीनियरिंग विभाग के डा. राजीव कुमार मिश्रा के नेतत्व में शोधकर्ताओं की एक ने इस अध्ययन को पूरा टीम किया। इसमें अहमदाबाद के भौतिक अनुसंधान प्रयोगशाला के प्रो. एस रॉमचंद्रन और दिल्ली प्रौद्योगिकी विश्वविद्यालय के एंड एकोस्टिक्स एडवांस एयर रिसर्च लेबोरेटरी के रिसर्च स्कालर कनगराज राजगोपाल शामिल रहे।



जीटी करनाल रोड पर मुकरबा चौक के पास उड़ती धूल= स्टीश कुमए यह अध्ययन सड़क किनारे के वातावरण में मौजूद नैनो कणों की सांद्रता की पहचान करने के उद्देश्य से किया गया है। इस शोध को डीटीयू के 'अर्बन क्लाइमेट जर्नल' में विस्तार से दिया गया है।

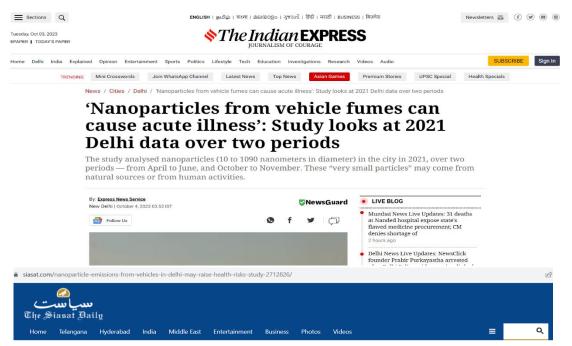
डा. राजीव कुमार मिश्रा ने बताया कि शोध में वर्ष 2021 में हुए लाकडाउन के दौरान और उसके बाद दिवाली के दौरान, उसके बाद की स्थिति को मुख्य रूप से अध्ययन किया गया है।

उद्देश्य उत्सर्जन स्रोतों के डसका आधार पर नैनो प्रदूषकों की कमी की पहचान करना था। इसमें सडक के किनारे के वातावरण, पैदल चलने वाले राहगीरों के मार्ग में मौजूद नैनो कणों को 10 से 1000 एनएम आकार तक मापा गया। दिल्ली में हवा के 10,000 से 10,00,000 घन सेंटीमीटर नैनोपार्टिकल पाए जाते हैं। इसमें ज्ञात हुआ कि लाक डाउन अवधि के दौरान जब वाहनों का चलना लगभग 50 प्रतिशत तक कम कर दिया गया था, तो नैनोपार्टिकल की एकाग्रता भी 31 प्रतिशत से कम हो गई थी। जबकि इसी बीच दिवाली के दौरान सामान्य उत्सर्जन की तुलना में नैनोपार्टिकल एकाग्रता को 35 प्रतिशत तक बढा दिया।

स्वास्थ्य के लिए अधिक खतरनाक हैं

नैनो पार्टिकल्स : शोध में यह पाया गया कि सीधे वाहन इंजन से गया कि साथ वाहन इजन स निकलने वाले 10 से 100 नैनो मीटर आकार के नैनोपार्टिकल सड़क के किनारे के क्षेत्रों में अधिक पाए गए। जब हवा की गति अधिक होती है, तो सडक के किनारे नैनो प्रदेषक संदक के आसपास के क्षेत्रों में फैल जाते हैं, जिससे सडक के पास रहने वाले निवासियों के लिए जोरिवम बढ् सकता है। नैनोपार्टिकल स्वास्थ्य के लिहाज से अधिक असुरक्षित हैं, क्योंकि यह पीएम 2.5 या पीएम 10 की तुलना मानव बाल के आकार से 600 गुना छोटे हैं, इसलिए फेफड़ों में प्रवेश कर सकते हैं। नैनो कणों में हमारे रक्तप्रवाह में प्रवेश करने की क्षमता है। यह मस्तिष्क के हिस्सों में जमा हो सकते हैं।



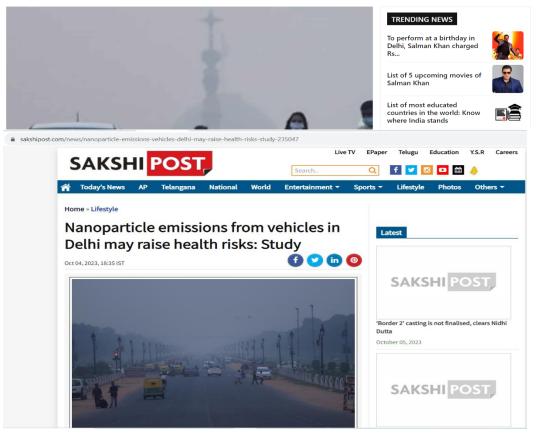


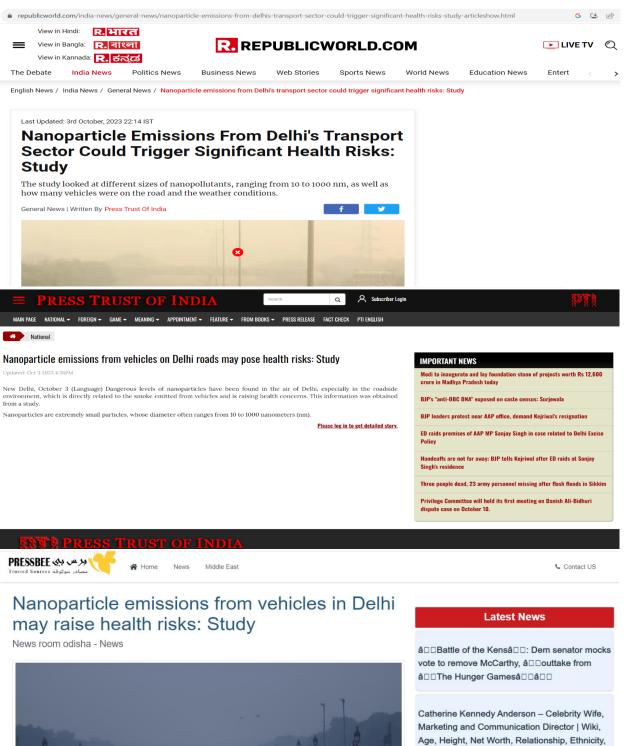
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## Nanoparticle emissions from vehicles in Delhi may raise health risks: Study

Due to rapid urbanisation, Delhi experiences frequent pollution events, and the particulate matter load exceeds the prescribed limit often.

IANS Indo-Asian News Service V | Posted by Mansoor Hameed | Published: 4th October 2023 8:24 pm IST



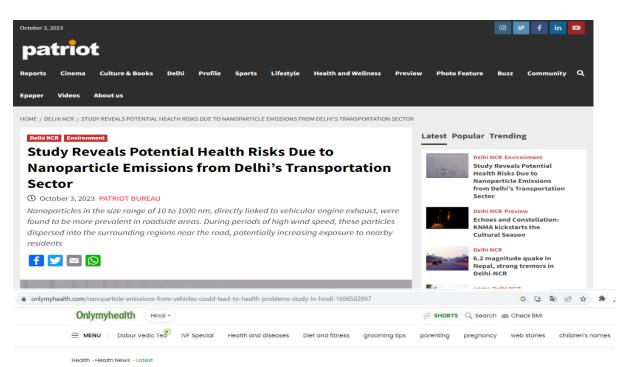


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## Study: Health problems increase due to exposure to nano particles emitted from vehicles.

According to a recent study, nano particles found in Delhi's vehicles can harm health. Let us know about it.



## Nanoparticle emissions from Delhi's transport sector could trigger significant health risks: Study



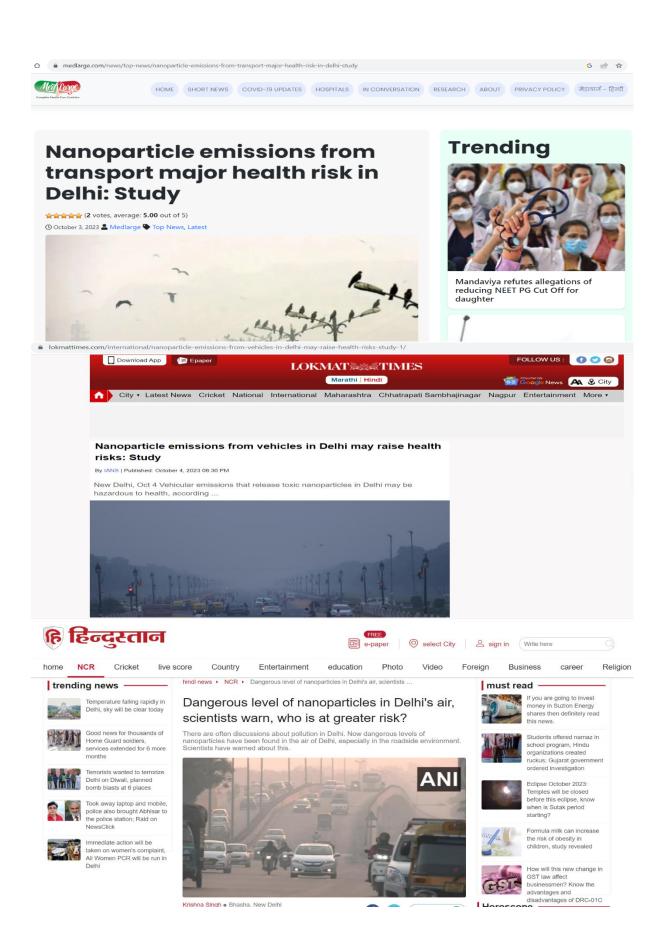
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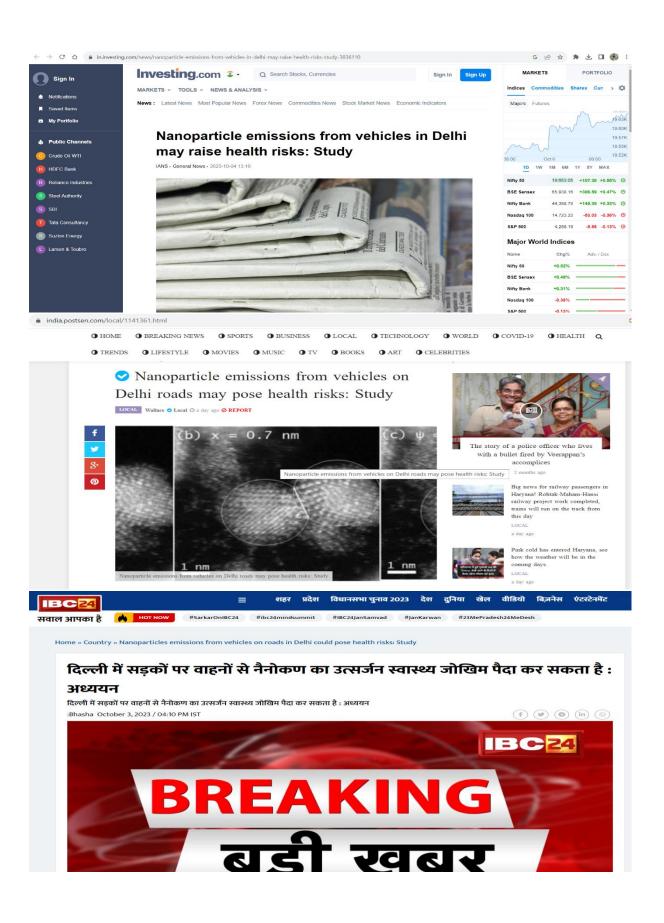
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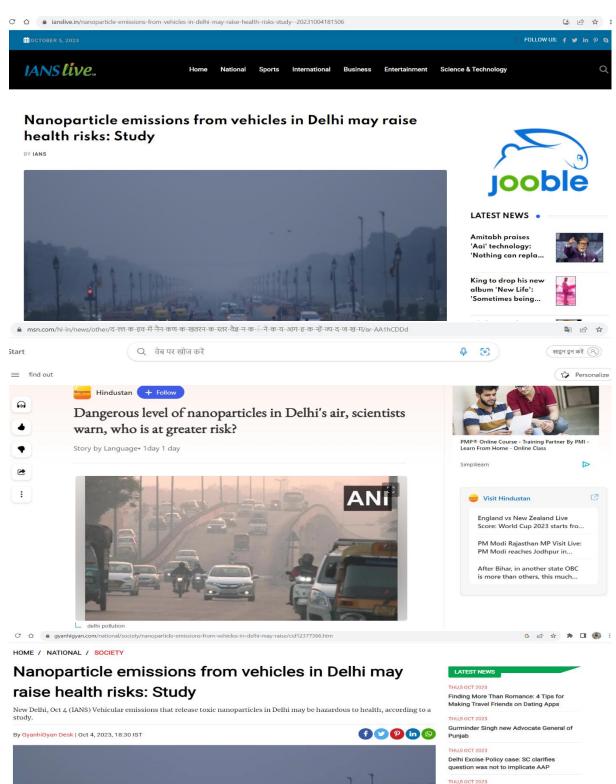












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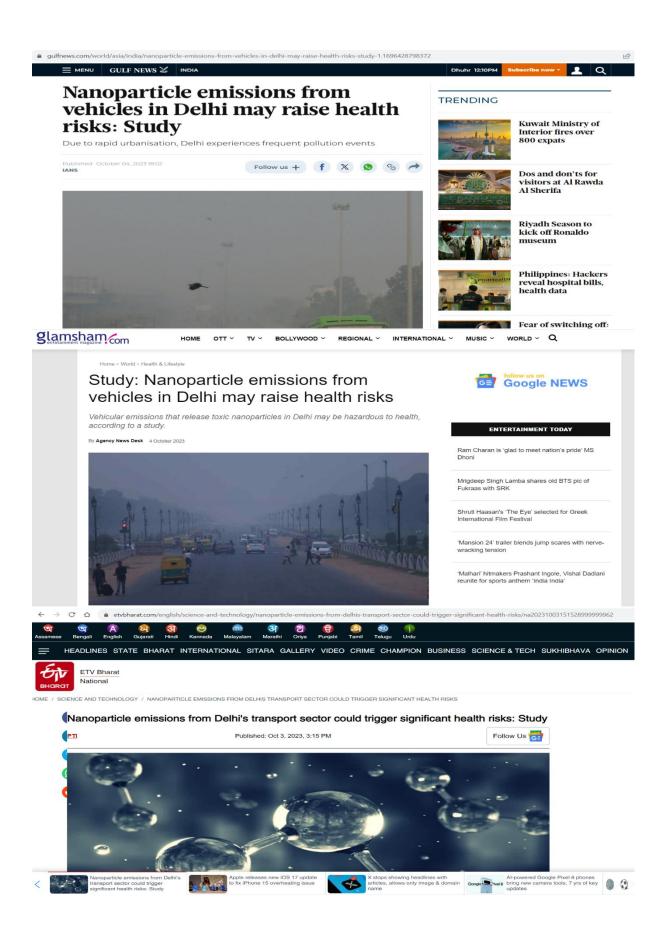
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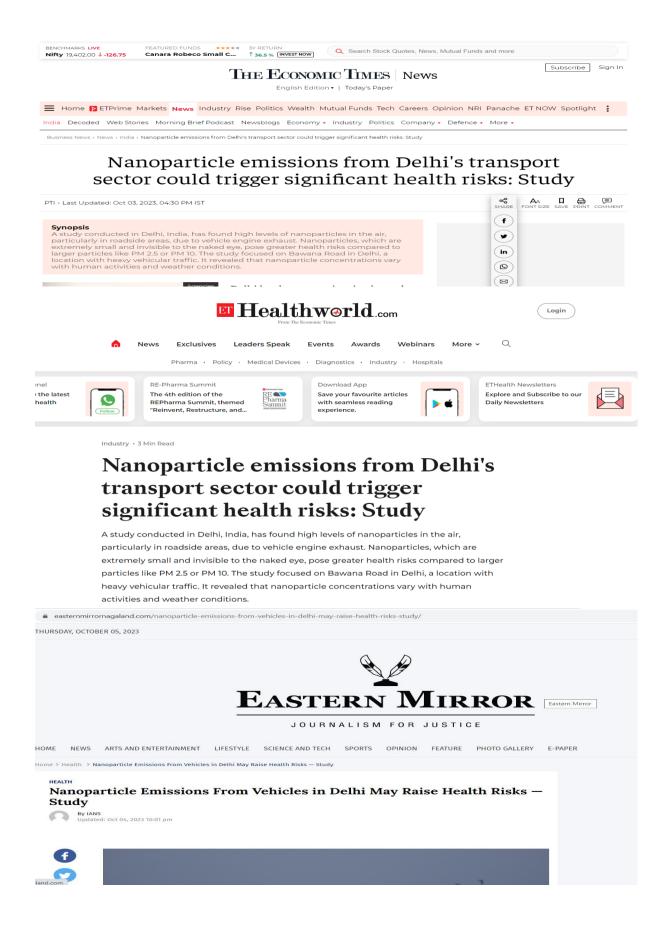
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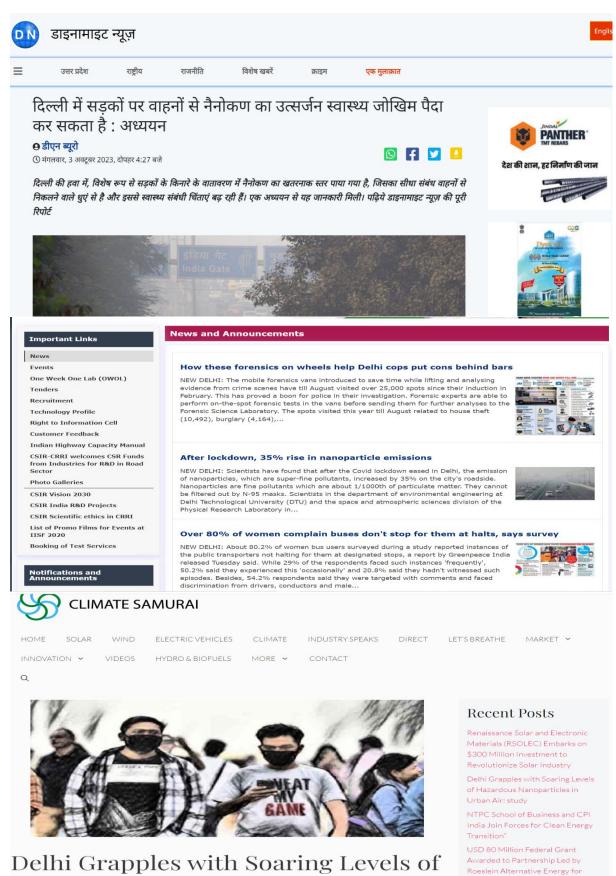
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