



Sensitivity of photochemical surface ozone formation regimes to emissions and meteorology in India

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Abstract. Atmospheric aerosols significantly contribute to air pollution and influence atmospheric chemistry, impacting air quality and public health. Decrease in aerosols can hinder the radical uptake sink of HO₂, and thus increase NO_x and OH, and subsequently increase ozone levels. This study investigates the seasonal variations of PM₁₀ and aerosol surface area and their effect on surface ozone levels in India, using the GEOS-Chem Chemical Transport Model for the years 2018 and 2022, two years with high and low simulated PM₁₀ concentrations, respectively. The results reveal substantial seasonal variations in PM₁₀ and aerosol surface area. In winter (DJF), higher PM₁₀ and aerosol surface area in the Indo-Gangetic Plain (IGP) and western Central India (CI) result from biomass burning and industrial activity, while coastal regions show lower aerosol surface area. A decrease in aerosol surface area is seen during the pre-monsoon (March–April–May; MAM) and monsoon (June–July–August–September; JJAS), followed by an increase in the post-monsoon (ON) season. As a result, aerosol-induced HO₂ uptake during winter and post-monsoon suppress surface ozone by approximately 5–10 μg m^{−3} in 2022 when compared to that of 2018. In contrast, during monsoon in 2022, the decrease in aerosol surface area caused an ozone increase of 5–7.5 μg m^{−3} when compared to that of 2018. On average, this increase in surface ozone due to the decrease in aerosols can be mitigated by reducing anthropogenic NO_x emissions by about 25 %–50 %. Thus, we recommend integrated strategies addressing aerosols, precursor emissions and regional meteorology to combat ozone pollution.

1 Introduction

In recent decades, surface ozone (O₃) has gained significant research attention due to its role as a transient (ranging from a few hours to a few weeks) secondary pollutant and its detrimental impacts on human health and agricultural yield (e.g., Mills et al., 2007; Avnery et al., 2011; Rathore et al., 2023; Gopikrishnan and Kuttippurath, 2024). It is an important trace gas in tropospheric chemistry, facilitating oxidation processes as the principal source of hydroxyl radicals (OH) (Logan, 1985) while also involved in chemical reactions with various organic molecules (Anderson et al., 2007). Ozone in the troposphere is a secondary pollutant generated via photochemical reactions that involve carbon monoxide (CO) and

volatile organic compounds (VOCs) together with nitrogen oxides (NO_x = NO + NO₂), recognised as ozone precursors (Crutzen et al., 1995; Gopikrishnan et al., 2022). However, these processes are dependent on solar radiation, with the reaction rate often peaking during summer months.

The formation dynamics of surface ozone is greatly affected by the presence of precursors such as NO_x and VOCs, resulting in NO_x-limited and VOC-limited regimes (Fiore et al., 2002; Lu et al., 2019). In the NO_x-limited regime, the concentration of NO_x is low compared to VOCs, implying that increasing NO_x levels can promote further ozone generation due to inadequate NO_x to completely react with VOCs. In a VOC-limited regime, the availability of VOCs is restricted, and further VOC emissions can significantly en-

hance ozone formation, whereas additional NO_x may have less effect on ozone formation. Nevertheless, these regimes are significantly affected by local emission sources, atmospheric chemistry and climatic conditions. For instance, aerosols can influence these regimes by modifying the concentration of hydroxyl (OH) and hydroperoxyl (HO_2) radicals, affecting the equilibrium between NO_x and VOCs and eventually determining the rate and extent of ozone generation in different locations (Jacob et al., 1995; Feng et al., 2016; Wang et al., 2018; Li et al., 2019a; Ivatt et al., 2022; Gopikrishnan et al., 2025).

Previous studies based on ground-based measurements from urban centers and industrial sites have shown substantial decrease in the particulate matter (PM) in the Indian cities post the implementation of National Clean Air Programme (NCAP), which primarily targets reducing the city-level PM pollution by 20 %–30 % in 131 cities when compared to that of 2017–2018. Gopikrishnan and Kuttippurath (2024) studied the Daily maximum 8 h average (MDA-8) ozone changes and observed that cities such as Visakhapatnam and Tirupati reported zero days of MDA-8 ozone surpassing 100 ppb post the implementation of NCAP in 2022. In contrast, during the 2018 base year, Visakhapatnam and Tirupati experienced about 60 and 10 such days, respectively. Nevertheless, some cities in the Indo-Gangetic Plains (IGP) of India, for instance, Agra, Singrauli, Ghaziabad, with decreasing aerosol loading and particulate matter pollution (Gopikrishnan and Kuttippurath, 2025), show an increase in the surface ozone levels. This trade-off between PM and O_3 was also mentioned in other highly populated regions of the world. For example, Wang et al. (2020) also observed that the opposite changes in surface ozone and $\text{PM}_{2.5}$ emerged as an unforeseen consequence of China's Clean Air Action Plan, aimed at reducing air pollution. Following the implementation of the plan, ozone levels increased over the summer in the North China Plain, due to reduced NO_x emissions and steady or rising VOC emissions, alongside significant decreases in $\text{PM}_{2.5}$ concentrations. This indicates that addressing ozone pollution requires deeper knowledge than broadly categorising it as NO_x -limited or VOC-limited, as the influence of PM and aerosols is significant (Zhao et al., 2023). Nevertheless, this variability in PM can also arise substantially due to changes in meteorology (Ivatt et al., 2022). Therefore, future air quality management systems worldwide must include the intricate relationship between O_3 , PM, precursor emissions and the regionally prevailing meteorology to effectively reduce both pollutants simultaneously.

Rural pollution is also a critical, yet often overlooked aspect of air quality management, especially in a country like India, where a significant portion of the population resides in rural areas (Pathak and Kuttippurath, 2022, 2024). Rural regions are not immune to air pollution; they face unique challenges such as solid fuel combustion for cooking, agricultural residue burning and increasing vehicular emissions due to expanding road networks (Bhuvaneshwari et al., 2019;

Chanana et al., 2023). These activities contribute to elevated levels of PM and ozone precursors, which can drift into urban areas, exacerbating pollution in cities through regional transport mechanisms. Furthermore, rural air pollution has profound implications for public health, as rural populations often lack access to healthcare infrastructure to address respiratory and cardiovascular diseases caused by poor air quality (Coker and Kizito, 2018; Manisalidis et al., 2020). It also impacts agricultural productivity, with rising ozone levels reducing crop yields, thereby threatening food security in agrarian economies like India (Pandya et al., 2022; Anagha et al., 2024). Addressing rural pollution is essential for achieving holistic improvements in air quality and ensuring equitable environmental health benefits across urban and rural populations.

Studies on the impact of aerosols on near-surface ozone formation have predominantly focused on the significance of Aerosol Optical Depth (AOD) extinction in photochemical processes that generate ozone (Bian et al., 2007; Kim et al., 2013; Wang et al., 2019a). AOD quantifies the degree to which aerosols scatter and absorb sunlight, hence influencing the quantity of solar radiation that reaches the Earth's surface. This subsequently affects the photochemical mechanisms that facilitate ozone synthesis in the lower atmosphere. Nevertheless, in areas with high aerosol concentrations, the reduction of surface ozone is not only governed by AOD extinction, but also the aerosol surface area by offering reactive sites for chemical reactions with molecules such as HO_2 (Macintyre and Evans, 2011; Song et al., 2021). The interaction between aerosols and these species leads to the elimination of ozone precursors, including OH radicals, which are vital for ozone production.

In this study, we aim to evaluate the impact of PM_{10} and aerosol surface area on surface ozone pollution in India. PM_{10} affects surface ozone formation mainly by enhancing aerosol optical depth and surface area, which reduce the photolysis rates of NO_2 and O_3 and decreases the concentrations of OH and HO_2 radicals, resulting in reduced net photochemical ozone production (Liu et al., 2024). To characterise the aerosol-driven ozone inhibition, we explicitly account for the surface area contributions from secondary organic aerosol, mineral dust, sea salt, black carbon, and primary organic carbon. As mineral dust and sea salt predominantly reside in the coarse fraction of PM_{10} , this framework captures aerosol–ozone interactions that are not fully represented by $\text{PM}_{2.5}$ alone (Bian and Zender, 2003; Bonasoni et al., 2004; George et al., 2017; Alves et al., 2018). Furthermore, given that national air-quality initiatives in India, such as the National Clean Air Programme (NCAP), target a 20 %–30 % reduction in PM_{10} relative to 2018 levels, this analysis also assesses the potential implications of PM_{10} mitigation for surface ozone pollution (NCAP, 2023). Since several studies are available on the air pollution concentrated in the urban regions of India, our analysis therefore explores the ozone formation regimes across entire India, using a high-resolution

GEOS-Chem Chemical Transport Model. This allows us to examine the impact of regional and larger scale dynamics of ozone production, which are often influenced by agricultural practices, biomass burning and long-range pollutant transport. By simulating two distinct years – 2018 (a high PM_{10} year and low rainfall year) and 2022 (a low PM_{10} year and normal rainfall year) with and without fixed meteorology for the two years – and incorporating/excluding the reactive absorption of HO_2 onto aerosols (via the γHO_2 parameter), we aim to understand how reductions in PM_{10} levels influence aerosol surface area and the secondary ozone chemistry (Figs. S1 and S2 in the Supplement). Furthermore, we categorize ozone generation regimes into NO_x -limited, VOC-limited and Aerosol Inhibited Regime (AIR) regions based on the dominant termination reactions in surface ozone formation for each model grid (Ivatt et al., 2022). Additionally, by scaling down NO_x emissions by 25 % and 50 %, we assess the efficacy of national and local emission control measures in mitigating NO_x emissions and their subsequent impact on surface ozone levels. This approach provides important insights into the chemistry between PM, precursor emissions and surface ozone across diverse geographical regions in India, highlighting the need for integrated air quality management strategies beyond urban boundaries.

2 Data and Methods

2.1 Region of Study

India can be delineated into six regions based on topographical characteristics, demographic factors, and pollution levels: Peninsular India (PI), Northwest (NW), Northeast (NE), Central India (CI), Indo-Gangetic Plain (IGP), and Hilly Regions (HR), as shown in Fig. 1. Peninsular India, which encompasses Kerala, Tamil Nadu, Karnataka, Goa, Andhra Pradesh, and Telangana, has three of the ten most populous states – Andhra Pradesh, Tamil Nadu, and Karnataka – and features substantial forest cover in the Western Ghats. Sources of pollution in this region are automobile emissions, industrial operations and biomass combustion, especially in urban areas like Bengaluru and Chennai. Central India, comprising Maharashtra, Madhya Pradesh, Chhattisgarh, Jharkhand and Odisha, is marked by low population density yet is linked to thermal power plants, coal mines and steel factories. Cities like Mumbai and Pune have higher pollution levels attributable to industrialisation and trash incineration. Northwest India includes Rajasthan and Gujarat, characterised by dry landscapes like the Thar Desert; pollution is attributed to dust from unpaved roads and construction activity, along with industrial emissions from sectors such as textiles. Northeast India comprises Tripura, Mizoram, Manipur, Nagaland, Meghalaya and Assam; despite its low population density and extensive vegetation, this region is affected by automobile pollution and slash-and-burn agriculture methods. The Indo-Gangetic Plain encompasses West

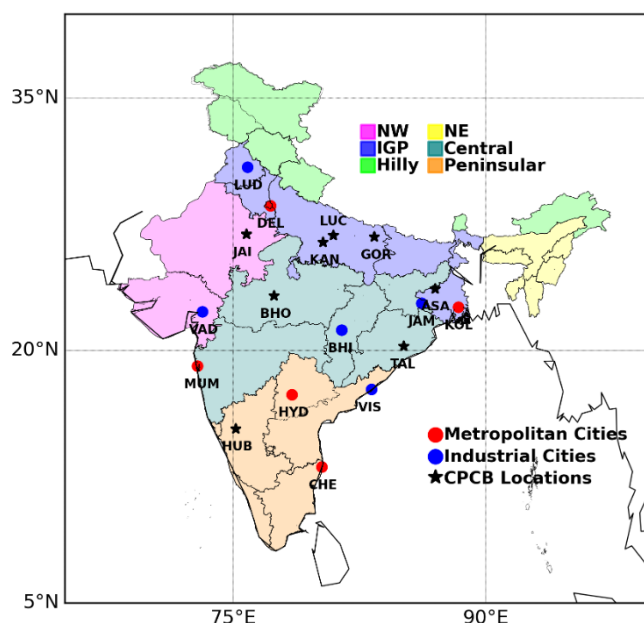


Figure 1. The region of study. The regions considered are also shown in the respective colours. Here, NW is Northwest, NE is Northeast India and IGP is Indo Gangetic Plain. The star marks indicate the respective ground-based stations. The red and blue dots indicate the metropolitan and Industrial cities in India. The abbreviations represent the first three letters of each city considered (e.g. DEL is Delhi), and their full names are given in Fig. S3.

Bengal, Bihar, Uttar Pradesh, Haryana and Punjab; it is characterised by a high population density exceeding 1000 individuals per square kilometre and has substantial pollution resulting from heavy traffic, industrial effluents and extensive crop residue burning during harvest periods. The Hilly Regions encompass Jammu and Kashmir, Himachal Pradesh, Uttarakhand, Sikkim and Arunachal Pradesh. These areas are less industrialised and ecologically sensitive, with economies dependent on agriculture and forestry. They experience localised pollution from vehicular emissions and biomass combustion, which is intensified by winter temperature inversions that confine pollutants near the surface (Kuttippurath et al., 2020; Gopikrishnan et al., 2022). A detailed summary of regional delineation, including constituent states and dominant pollution characteristics, is provided in Table S1 in the Supplement.

2.2 GEOS-Chem

GEOS-Chem is an extensive model developed for modelling complex oxidant-aerosol chemistry in the troposphere and stratosphere (Zhang et al., 2011; Kim et al., 2015). We have used GEOS-Chem version 14.4.3 in this study (<https://doi.org/10.5281/zenodo.13314490>, The International GEOS-Chem User Community, 2024). The model complies with the most recent JPL/IUPAC guidelines for

chemical mechanisms, with substantial modifications that improve the depiction of diverse chemical processes, including those related to isoprene, aromatics and nitrates (Bates et al., 2023). Recent advancements have enhanced the treatment of complex chemical reactions, including methanol synthesis and mercury redox chemistry, facilitating more precise atmospheric forecasts (Horowitz et al., 2017; Guil-López et al., 2019). Additionally, the model incorporates the reactive absorption of NO_x by aerosols and computes aerosol hygroscopicity, which is essential for comprehending aerosol-cloud interactions. GEOS-Chem offers a comprehensive framework for investigating atmospheric chemistry and its effects on air quality and climate change, accommodating various chemical species and reactions, including halogens and hydroxymethanesulfonate. This model provides thorough understanding into ozone generation mechanisms and pollutant interactions, serving as an essential resource for addressing air quality challenges and comprehending the wider implications of atmospheric dynamics.

GEOS-Chem has been rigorously validated by researchers globally. For instance, Travis and Jacob (2019) showed that, despite the successful simulation of ozone and its precursors in the SEAC4RS aircraft data below 1 km of altitude, MDA8 surface ozone was biased high in the model by $+6\mu\text{g m}^{-3}$ on average. David et al. (2019) used the GEOS-Chem transport model in India and observed that the model reasonably simulated the tropospheric O_3 abundances and vertical profiles, with a mean bias of 1–3 DU compared to observations for the period 2000–2015. Christiansen et al. (2022) found that the GEOS-Chem model, when validated against ozonesonde, aircraft, and satellite observations globally, showed reasonable agreement with a mean bias of 1–3 DU for the period from 1990–2017. Mao et al. (2024) reported that the GEOS-Chem model, validated against MDA8 O_3 concentration observations in central and eastern China from May–July 2017, showed a strong correlation of 0.77 (95 % confidence level). In Nanjing, the simulated MDA8 O_3 concentrations converged with the observed trend, with a correlation coefficient of 0.65, a normalised mean bias (NMB) of 5 %, and a normalised mean error (NME) of 21 %. Karambelas et al. (2022) used the GEOS-Chem model to simulate $\text{PM}_{2.5}$ concentrations during high PM episodes in the fall of 2015 and 2017, finding that the model underestimated observed concentrations by 28 %, with an average observed $\text{PM}_{2.5}$ concentration of $142\mu\text{g m}^{-3}$ compared to $125\mu\text{g m}^{-3}$ from the model. Pai et al. (2022) validated the GEOS-Chem simulation against speciated airborne aerosol measurements over India and reported significantly reduced normalized mean biases: $\text{NMB}_{\text{Modified}} = 0.19$ compared to $\text{NMB}_{\text{Base}} = 0.61$ overall – with nitrate improving from 1.64–0.08 and ammonium from 0.90–0.49. Despite this underestimation, the model demonstrated strong spatial correlations with observed concentrations ($r^2 = 0.67$), accurately capturing the spatial distribution of $\text{PM}_{2.5}$ across India. David and Ravishankara (2019) used the GEOS-Chem

model to estimate boundary layer ozone contributions across eight regions of the Indian subcontinent, finding that boundary layer ozone in northern India, Pakistan, and Sri Lanka is largely (about 65 %–70 %) influenced by regions outside the subcontinent. They also highlighted the growing importance of central India in contributing to ozone pollution, with regional meteorology playing a key role in the redistribution of boundary layer ozone and its precursors including Chlorine (Cl).

In this study, GEOS-Chem is simulated in a $2.0^\circ \times 2.5^\circ$ latitude by longitude grid resolution globally for the boundary conditions for the years 2018 and 2022. One year spin up simulations are used to generate the model restart files and the boundary conditions. The nested model has a grid resolution of $0.25^\circ \times 0.3125^\circ$ latitude by longitude and a spin up period of 1 month. We use the data from the lowest level of the ETA coordinate system from GEOS-Chem model simulations to analyse the correlation between modelled and observed PM_{10} and ozone data. The model is driven by the GEOS-FP assimilated meteorological data provided by the Goddard Earth Observing System (GEOS) of the Earth Modelling and Assimilation Office (GMAO) and with a temporal resolution of 1 h (Lucchesi, 2013). Emissions in the GEOS-Chem model are handled using the Harvard–NASA Emission Component (HEMCO; Keller et al., 2014). Anthropogenic emissions are derived from the Community Emissions Data System (CEDS), which provides global coverage at a native resolution of $0.1^\circ \times 0.1^\circ$, with data aggregated to the model resolution (Hoesly et al., 2018). Sea salt emissions are calculated offline following the approach of Zhu et al. (2019), while dust emissions are calculated offline following the method of Ginoux et al. (2012). Emissions of biogenic origin are simulated using the Model of Emissions of Gases and Aerosols from Nature (MEGAN), which provides global emissions at 0.01° latitude $\times$ 0.01° longitude resolution (Guenther et al., 2012) and Soil NO_x emissions follow the approach outlined by Hudman et al. (2012). Biomass burning emissions are sourced from the Global Fire Emissions Database version 4 (GFED4), which provides a global coverage at 0.25° latitude $\times$ 0.25° longitude (Giglio et al., 2013).

GEOS-Chem uses a set of chemical mechanisms implemented with the Kinetic PreProcessor (KPP) (Damian et al., 2002). The standard chemical mechanism in GEOS-Chem has undergone continuous development since the original tropospheric gas-phase scheme of Bey et al. (2001), with subsequent extensions to include aerosol chemistry (Park, 2004), stratospheric chemistry (Eastham et al., 2014), and a detailed tropospheric–stratospheric halogen chemistry module (Wang et al., 2019b), which include 299 chemical species (Fritz et al., 2022). Sulfate–nitrate–ammonium aerosols are simulated with a bulk thermodynamic approach (Park, 2004), where gas–particle partitioning is calculated using ISOR-ROPIA II (Fountoukis and Nenes, 2007). Sea salt is represented with two modes distinguishing fine and coarse parti-

cles (Jaeglé et al., 2011), while mineral dust is described using four particle size bins (Fairlie et al., 2007). Furthermore, the simple SOA scheme is employed for secondary organic aerosols, based on reversible partitioning of semivolatile products of VOC oxidation (Pye et al., 2010). Furthermore, in GEOS-Chem, PM_{10} is diagnosed as the sum of $\text{PM}_{2.5}$ plus size-resolved dust (70 % of DST2, all of DST3, 90 % of DST4) and coarse sea-salt (SALC scaled by the hygroscopic growth factor, e.g., 1.86 at 35 %–50 % RH), with the final values archived at standard temperature and pressure (STP) using the ideal gas law (GEOS-Chem Support Team, 2023).

Furthermore, we use ground-based measurements from Central Pollution Control Board (CPCB) monitoring sites situated in different regions of India, as shown in Figs. 1 and S3, reflecting a wide array of meteorological conditions. The CPCB monitors ambient ozone using UV photometric methods (based on the absorption of UV light by ozone at 254 nm), chemiluminescence (where ozone reacts with a reagent to produce light that is quantified), and chemical methods (colorimetric reactions providing concentration estimates). PM_{10} is measured using gravimetric analysis (filter-based mass determination after sampling), Tapered Element Oscillating Microbalance or TEOM (continuous real-time mass measurement through frequency change of a vibrating element), and beta attenuation techniques (attenuation of beta radiation as it passes through particle-laden filters to infer mass concentration) and these data are available at a resolution of 1 h (NAAQS, 2023).

The altitude of the lowest model level may not consistently align with the actual altitude of ground-based monitoring locations, thereby causing inconsistencies between modelled and observed data. Therefore, we calculate the correlation coefficient for the time series of ozone and PM_{10} values at each station to evaluate the strength of the association between these variables, as shown in Fig. S3. The correlation coefficient between modelled and observed values of PM_{10} and ozone is above 0.6 for most of the stations, except in Jaipur in the northwest, where the correlations are 0.5 and 0.4 for PM_{10} and ozone, respectively. Although these moderate to strong correlations suggest a satisfactory alignment between the model and the facts, certain systematic differences remain. The GEOS-Chem model generally underestimates PM_{10} concentration, possibly attributable to its $0.25^\circ \times 0.3125^\circ$ resolution, which fails to account for fine-scale local variations, including point sources of pollution. In contrast, the model often overestimates surface ozone concentrations for most of the stations considered here. Despite the model's higher bias caused by uncertainties in emissions of anthropogenic and biogenic precursors, uncertainties in ozone sinks such as dry deposition, and the coarse resolution of the model limits its ability to capture ozone titration and other localised chemical processes (Westervelt et al., 2019). Our analysis focuses on the differences between simulations for 2018 and 2022, with model performance evaluated using ground-based CPCB observations of surface

ozone and PM_{10} (Figs. 2, S4 and S5). GEOS-Chem broadly captures regional patterns of annually averaged PM_{10} and ozone, but ground-based stations show higher PM_{10} in north west and IGP ($140\text{--}160\text{ }\mu\text{g m}^{-3}$ vs. model $120\text{--}140\text{ }\mu\text{g m}^{-3}$, which is about 15 %–20 % low bias) and lower values in central and peninsular India (about 10 %–15 % high bias). For ozone, the model underestimates HR and IGP peaks by about 10 %–15 % (observed $> 150\text{ }\mu\text{g m}^{-3}$) while overestimating coastal peninsular levels by about 10 %–20 % (observed $70\text{--}90\text{ }\mu\text{g m}^{-3}$). The model reproduces the temporal patterns reasonably well, with annually averaged differences between modelled and observed values remaining within 10 %–20 % across most stations. Despite a few regionally specific biases (higher relative differences in PM_{10} in cities like Talcher, Noida, Jodhpur during the monsoon season), the consistent performance across two meteorologically different years lends confidence to the overall robustness of the high-resolution GEOS-Chem simulations for India.

2.3 Methods

The aerosol uptake coefficient, which quantifies the efficacy of particles in absorbing reactive gases, substantially impacts surface ozone concentrations (Jacob, 2000). In areas with elevated aerosol concentrations, aerosols possessing a high uptake coefficient can more efficiently eliminate ozone precursors, including HO_2 and other reactive radicals, from the atmosphere (Li et al., 2018). These aerosols interact with atmospheric chemical species, reducing the levels of OH and HO_2 radicals, which are essential for ozone synthesis. The aerosol uptake coefficient of HO_2 (γ_{HO_2}), however, is affected by several factors, including the physical and chemical characteristics of the aerosol particles, such as size, composition and moisture content (George and Abbatt, 2010). These parameters show the efficacy of aerosol interactions with atmospheric radicals, thereby affecting ozone production.

Termination rates of chain reactions were determined using archived species concentrations and physical parameters such as temperature, pressure and humidity, in addition to aerosol properties, as in Ivatt et al. (2022). Although we do not classify radical product generation in peroxy-radical self-reactions as termination stages, we regard non-radical products as ongoing termination processes. The heterogeneous loss rate of HO_2 was assessed by evaluating the radius and surface area of different aerosol forms from the archived model outputs. For our simulations, we employed a default HO_2 reactive uptake coefficient (γ_{HO_2}) of 0.2. Laboratory studies of pure synthetic aerosols indicate lower uptake coefficients ($\gamma_{\text{HO}_2} < 0.2$), whereas real-world aerosol studies reveal values between 0.08–0.40, implying that elements such as transition metals may increase aerosol uptake (Kolb et al., 2010; Christian et al., 2018). We have also simulated the model with γ_{HO_2} to be zero to evaluate its impact on ozone formation mechanism. While we presumed H_2O to be the exclusive product of HO_2 absorption, our results remain

valid when H_2O_2 is considered. Employing a singular γHO_2 value likely oversimplifies the variability, as existing models fail to account for its temporal oscillations (Stavrakou et al., 2013; Sheehy et al., 2010).

The heterogeneous radical loss rate of HO_2 is determined by iterating through the radius and surface area of various aerosol types, which includes sulfate–nitrate–ammonium thermodynamics, secondary organic aerosol, dust, sea salt, black carbon, and primary organic carbon (Ivatt et al., 2022). These factors, along with temperature and air density, are used to calculate the first-order loss rate of HO_2 . Furthermore, emissions and photochemical reactions are strongly affected by seasonal variations in India. The seasons are defined as Winter (December–January–February; DJF), Pre-monsoon (March–April–May; MAM), Monsoon (June–July–August–September; JJAS) and Post-monsoon (October–November; ON).

3 Results

3.1 Distribution of Surface Ozone and PM for the years 2018 and 2022

3.1.1 Annual Distribution

Figure 2 shows the annual average PM_{10} and ozone in India for the years 2018 and 2022. The highest PM_{10} concentrations ($130\text{--}150\mu\text{g m}^{-3}$) occur in the north west dry regions due to dust mobilisation from the Thar Desert (Santra et al., 2018; Kashyap et al., 2025), consistent with studies showing dry and semi-arid areas as key mineral dust sources (Maji and Sonwani, 2022). Ground-based observations here often exceed $160\mu\text{g m}^{-3}$, indicating that the model underestimates by about 10 %–20 %. This bias may partly arise from uncertainties in dust emission parameterisations and coarse-resolution representation of wind-driven dust mobilisation in GEOS-Chem (Saxena and Pandey, 2025). IGP also shows higher PM_{10} ($120\text{--}140\mu\text{g m}^{-3}$) from industrial, vehicular and agricultural emissions including biomass and crop-residue burning (Devi et al., 2020, 2024; Hassan et al., 2023). Station data shows average concentrations of about $140\text{--}160\mu\text{g m}^{-3}$, indicating that the model is biased low by about 15 %. Such biases are likely linked to uncertainties in anthropogenic emission inventories, particularly for primary PM, NO_x and VOC emissions in India, which are often under-reported or lack temporal and sectoral detail (Smith et al., 2001; Ge et al., 2024; Tripathi et al., 2025). In addition, simplifications in secondary aerosol formation pathways can further contribute to PM underestimation (Lane et al., 2008). Despite higher emissions, greater humidity and rainfall promote wet deposition and particle scavenging, lowering levels relative to north west India (Sen et al., 2017; Singh et al., 2023). Central India records $60\text{--}100\mu\text{g m}^{-3}$ with industrial hotspots up to $130\text{--}150\mu\text{g m}^{-3}$ (Lokhande and Khan, 2021; Saini et al., 2023), but observations show about 50--

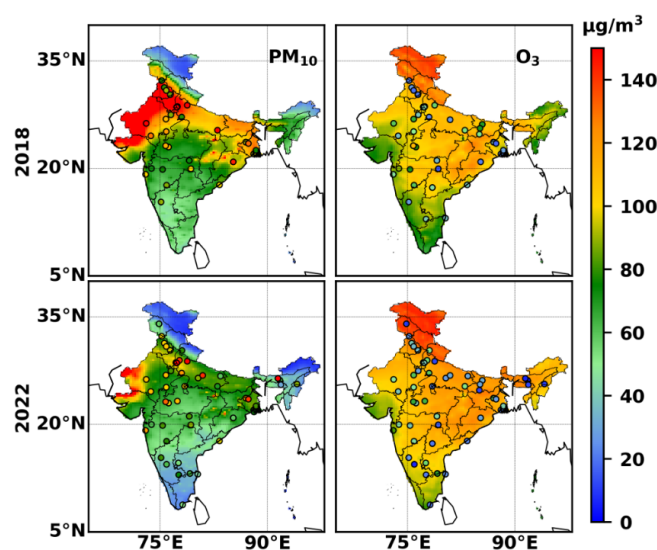


Figure 2. Annual average distribution of PM_{10} and ozone for the years (top) 2018 and (bottom) 2022 from GEOS-Chem simulations. The dots show the annually averaged mean values of PM_{10} and ozone at the respective ground-based CPCB stations for the years 2018 and 2022.

$90\mu\text{g m}^{-3}$, suggesting the model slightly overestimates by about 10 %–15 %. North east India shows lower levels ($40\text{--}70\mu\text{g m}^{-3}$) due to sparse population and vegetation sinks (Guttikunda and Nishadh, 2022; Kumari et al., 2025).

Surface ozone peaks in the hilly regions ($120\text{--}140\mu\text{g m}^{-3}$) due to high stratospheric-to-troposphere transport (STT). Ground-based data here often reach $150\text{--}160\mu\text{g m}^{-3}$, showing that model underpredicts by about 15 % (Phanikumar et al., 2017; Wang et al., 2024a). STT is included in the GEOS Chem model through the unified tropospheric–stratospheric chemical mechanism (UCX; Eastham et al., 2014). High values also occur in IGP and eastern central India ($100\text{--}120\mu\text{g m}^{-3}$) driven by secondary photochemistry from NO_x and VOCs (Payra et al., 2022; Sinha and Sinha, 2019; Mahilang et al., 2021; Kuttippurath et al., 2022). These enhancements occur under stagnant conditions, where low wind speeds ($< 3.2\text{ m s}^{-1}$), suppressed boundary-layer mixing, and limited dispersion enhance the accumulation of ozone and its precursors near the surface (Gopikrishnan and Kuttippurath, 2024). Observations in these regions are typically $110\text{--}130\mu\text{g m}^{-3}$, reflecting a model underestimation of about 10 %–15 %. Peninsular India shows the lowest ozone ($80\text{--}100\mu\text{g m}^{-3}$) due to coastal ventilation, humidity, precipitation and oceanic loss, with observed levels closer to $70\text{--}90\mu\text{g m}^{-3}$, showing that the model overestimates by about 10 %–20 %. This positive bias likely arises from uncertainties in NO_x and VOC emission inventories in peninsular India and an underrepresentation of coastal ozone loss processes, including marine deposition and halogen chemistry, in CTMs (Li et al., 2019b; Lakshmi et al., 2024; Galbally and Roy,

1980; Lavanyaa et al., 2023; Nilaya et al., 2024). A detailed discussion of seasonal changes in surface ozone is provided in Sect. S1.1 in the Supplement.

Seasonally, PM_{10} is highest in IGP during winter and post-monsoon ($100\text{--}140\text{ }\mu\text{g m}^{-3}$) under stable, inversion conditions and intensified biomass and crop-residue burning (Mhawish et al., 2020; Paulot et al., 2022; Jayachandran and Rao, 2024; Payra et al., 2022; Mogno et al., 2021). The north west records its peak during the monsoon from dust storms (Yadav et al., 2022; Yu et al., 2023), while pre-monsoon sees widespread elevated PM_{10} ($60\text{--}100\text{ }\mu\text{g m}^{-3}$) with north west reaching $140\text{--}160\text{ }\mu\text{g m}^{-3}$ due to dust, traffic, industrial and construction sources (Patil et al., 2013; Garg and Gupta, 2020). These high levels pose severe health risks, particularly for vulnerable groups (Adhikary et al., 2024; Pathak et al., 2024; Gopikrishnan and Kuttippurath, 2025). Ozone peaks in pre-monsoon, especially over central, IGP and north east India, where it exceeds $140\text{--}160\text{ }\mu\text{g m}^{-3}$, driven by strong solar radiation and abundant precursors (Kutal et al., 2022; Sinha et al., 2014). Winter ozone is also high in central India ($100\text{--}120\text{ }\mu\text{g m}^{-3}$), followed by IGP and north east India ($80\text{--}100\text{ }\mu\text{g m}^{-3}$) under stagnant meteorology and local emissions (David and Nair, 2011; Gao et al., 2020). In contrast, monsoon levels are lowest, with the west coast at $60\text{--}80\text{ }\mu\text{g m}^{-3}$, north west at $80\text{--}100\text{ }\mu\text{g m}^{-3}$ and north east and IGP at $120\text{--}140\text{ }\mu\text{g m}^{-3}$, owing to cloud cover, precipitation and circulation that suppress ozone production and enhance removal (Lal et al., 2014; Lu et al., 2018). Post-monsoon levels are moderate ($100\text{--}120\text{ }\mu\text{g m}^{-3}$) across central, north west and IGP, with peninsular and north east India slightly lower ($60\text{--}90\text{ }\mu\text{g m}^{-3}$) (Lu et al., 2018). A detailed discussion of seasonal changes in surface ozone is provided in Sect. S1.2.

3.1.2 Comparison of PM_{10} and Ozone levels

The relationship between PM_{10} and ozone is often inverse, attributable to their distinct formation mechanisms and atmospheric interactions, but varies substantially depending on meteorological conditions and precursor availability (Jia et al., 2017). Higher PM levels, frequent during winter in areas such as IGP, often correlate with lower ozone concentrations (Song et al., 2022). PM disperses and absorbs sunlight, reducing the solar radiation essential for photochemical ozone formation, which depends on NO_x and VOCs concentration in the presence of intense sunlight. This reduction in actinic flux directly suppresses photolysis frequencies (e.g., $J(\text{NO}_2)$ and $J(\text{O}^1D)$), thereby limiting radical production (OH and HO_2) and slowing ozone formation rates (Harrison, 2007). Black carbon (BC), a constituent of PM from combustion, can directly scavenge ozone by acting as a deposition surface. This occurs through heterogeneous uptake of O_3 onto the porous and highly reactive BC surface, where ozone is adsorbed and decomposed into molecular oxygen, effectively removing O_3 from the gas phase (Gao et al., 2018). However, this direct BC–ozone scavenging pathway

is not explicitly represented in most chemical transport models (Koch et al., 2009), but its overall contribution is considered modest compared to dominant photochemical production and deposition processes. The large BC loadings and high surface-area densities in South Asia may render this pathway locally important, particularly under stagnant wintertime conditions (Kumar et al., 2015; Pathak et al., 2024). Furthermore, stable atmospheric conditions including temperature inversions that confine PM at the surface also impede vertical mixing, limiting ozone transport from the upper atmosphere, hence facilitating the accumulation of ozone near the surface (Gopikrishnan and Kuttippurath, 2025). In addition, high PM loading enhances aerosol radiative cooling at the surface, reducing sensible heat flux and suppressing turbulent mixing, which leads to a shallower planetary boundary layer (PBL; Budakoti and Singh, 2021). In such cases, the co-accumulation of PM and ozone near the surface does not imply a simple inverse relationship, as aerosol–radiation interactions can further suppress the PBL height (Li et al., 2017; Zou et al., 2017), thereby enhancing pollutant trapping. A shallower PBL reduces vertical dilution, increasing the residence time of both ozone and its precursors in the near-surface layer (Torres-Vazquez et al., 2022). Under suppressed PBL conditions and high NO_x emissions, ozone titration can also occur, leading to additional complexity in the surface ozone response (Lin et al., 2008; Li et al., 2024). Conversely, cleaner environments with reduced PM concentrations enhance ozone generation due to the lack of light scattering and scavenging, which fosters photochemical activity (Rathore et al., 2023; Sicard et al., 2023). Though GEOS-Chem robustly captures aerosol-induced photolysis attenuation and radical scavenging (HO_2 uptake), the direct heterogeneous loss of ozone on black carbon surfaces remains a source of model uncertainty as it is not explicitly represented in the standard chemical mechanism (Wang et al., 2022).

Figure 3 shows the change in PM_{10} and ozone in India for the year 2022 when compared to that of 2018. In the monsoon and post-monsoon seasons, PM concentrations drop by $40\text{--}60\text{ }\mu\text{g m}^{-3}$ in the north west, IGP and eastern central India, while reductions of $10\text{--}30\text{ }\mu\text{g m}^{-3}$ occur in other regions, due to changes in the spatial distribution of precipitation as shown in Figs. S6 and S7. Such reductions in PM weaken aerosol direct radiative effects by decreasing scattering and absorption of incoming shortwave radiation, thereby increasing surface solar irradiance and actinic flux, thereby altering photolysis rates (Singh et al., 2023). Figure S8 presents the changes in aerosol surface area between 2018 and 2022, whereas Fig. S9 illustrates the contributions from different aerosol sources in these two years. The reduction in PM facilitates increased solar radiation penetration (Aladwani et al., 2024), hence promoting photochemical ozone production, resulting in a substantial rise in ozone concentrations ($20\text{--}30\text{ }\mu\text{g m}^{-3}$ in most of India during monsoon and $20\text{--}30\text{ }\mu\text{g m}^{-3}$ in peninsular and western central In-

dia during post-monsoon). This response is primarily driven by aerosol direct effects, through enhanced photolysis rates of NO_2 and O_3 , while reduced aerosol loading may also weaken aerosol–cloud interactions that otherwise increase cloud albedo and lifetime during the monsoon (Singh et al., 2023). The decrease in PM_{10} reduces the surface area accessible for heterogeneous processes that might otherwise reduce ozone levels (Gopikrishnan et al., 2025). Conversely, in pre-monsoon and winter seasons, PM concentrations increase by $10\text{--}30\text{ }\mu\text{g m}^{-3}$ regionally, particularly in north west and central IGP, inhibiting ozone formation due to reduced solar radiation and enhanced heterogeneous chemistry that depletes reactive species such as HO_2 (Wang et al., 2024b). In these seasons, enhanced aerosol direct effects dominate by attenuating surface radiation, while aerosol indirect effects, through increased cloud condensation nuclei and persistent low-level clouds, can further suppress photolysis, especially under high-humidity conditions (Wu et al., 2020). The stagnant meteorological conditions associated with winter (DJF) and post-monsoon season (ON), including temperature inversions, elevated humidity, and minimal wind speeds, intensify PM buildup, particularly in IGP (Fig. S10), resulting in further reductions in ozone levels (Gopikrishnan and Kutippurath, 2024). Elevated PM loading under such conditions also enhances surface radiative cooling, suppressing turbulent mixing and leading to a shallower planetary boundary layer, which reinforces pollutant trapping (Torres-Vazquez et al., 2022). Nonetheless, geographical disparities remain, exemplified by a $20\text{ }\mu\text{g m}^{-3}$ reduction in PM in the Upper Northeast of IGP during pre-monsoon, which causes a $20\text{--}30\text{ }\mu\text{g m}^{-3}$ rise in ozone levels, and a $10\text{--}20\text{ }\mu\text{g m}^{-3}$ decline in PM near the west coast, resulting in a $40\text{--}60\text{ }\mu\text{g m}^{-3}$ increase in ozone. In winter, ozone variations are negligible (within $10\text{ }\mu\text{g m}^{-3}$) over much of India; however, a reduction in PM ($10\text{--}30\text{ }\mu\text{g m}^{-3}$) in the northeast correlates with a $20\text{--}30\text{ }\mu\text{g m}^{-3}$ rise in ozone levels. The seasonal variability is additionally affected by the aerosol–photolysis feedback, wherein PM influences ozone by absorbing important species such as odd-hydrogen radical family (H_xO_y , including OH, HO_2 , and related peroxides) and NO_x , hence reducing surface photolysis rates and inhibiting ozone production (Ivatt et al., 2022; Gopikrishnan et al., 2025).

4 Discussion

4.1 Impact of HO_2 uptake onto aerosol on surface ozone

The GEOS-Chem model simulations for 2018 and 2022 were performed without and with aerosol uptake of HO_2 to study the impact of γHO_2 on surface ozone concentrations in India in the absence of this mechanism. The results show that the aerosol absorption of HO_2 substantially influences ozone concentrations, as shown in Fig. 4, especially in areas like the IGP and eastern central India. The increase in ozone con-

centrations in these regions without aerosols (Fig. S11), attributed to the absorption of HO_2 by aerosols, varies from $5\text{--}10\text{ }\mu\text{g m}^{-3}$, with the most significant changes during winter and post monsoon season. Winter season is defined by higher aerosol surface area, which promotes heterogeneous processes that enhance the removal of HO_2 , hence restricting its availability for ozone formation (Dyson et al., 2022). In 2022, aerosol concentrations rose substantially in winter when compared to 2018, especially in the middle IGP and western central India. This augmentation of aerosol surface area in 2022 inhibited ozone generation by about $5\text{--}7.5\text{ }\mu\text{g m}^{-3}$ in these areas. The most notable changes in ozone concentrations were during post-monsoon, with an increase of about $5\text{--}10\text{ }\mu\text{g m}^{-3}$ when no aerosol absorption is considered, especially in IGP and western central India. In 2022, there was a rise in aerosol surface area in western central India compared to 2018, resulting in an additional suppression of surface ozone. The pre-monsoon season period had similar changes in ozone concentration, with an increase of about $2.5\text{--}5\text{ }\mu\text{g m}^{-3}$, whereas the monsoon period observed minimal variation, with ozone concentration changes between 0 and $5\text{ }\mu\text{g m}^{-3}$. Thus, the changes in ozone concentrations are directly influenced by aerosol surface area, which amplifies the heterogeneous elimination of HO_2 , thereby restricting its availability for ozone synthesis (Anurose et al., 2024). Seasonal fluctuations in aerosol load, influenced by factors like monsoon dynamics, atmospheric transport, and emissions, greatly influence the effects of aerosol– HO_2 interactions on ozone levels. The rise in aerosol concentrations in 2022, especially during winter and post-monsoon, can be attributed to higher anthropogenic activities increasing sulphate, BC and OC, which enhance the suppression of ozone production in these seasons (Austin et al., 2015; Zhang et al., 2023).

In addition to the simulation with a γHO_2 of 0.2, a sensitivity analysis using $\gamma\text{HO}_2 = 0.1$ was conducted for the years 2018 and 2022, as shown in Figs. 4 and S12, respectively. This lower value was chosen, as previous studies suggest that $\gamma\text{HO}_2 = 0.2$ may overestimate heterogeneous uptake, particularly over East Asia, where values closer to 0.1 have been reported under typical ambient conditions (Yang et al., 2023). Given the large uncertainty and variability in γHO_2 arising from factors such as aerosol composition, relative humidity, and particle acidity (Lakey et al., 2024), it is critical to assess the sensitivity of modeled ozone to a range of uptake efficiencies. This uncertainty can lead to substantial differences in simulated ozone, as even a halving of γHO_2 (from 0.2–0.1) significantly reduces the magnitude and spatial extent of ozone suppression. The comparison between the two scenarios indicates that the impact on surface ozone concentrations is both spatially and seasonally dependent, with $\gamma\text{HO}_2 = 0.2$ yielding a more pronounced and widespread increase when compared to a no aerosol uptake scenario. During the DJF and ON seasons, the increase in ozone is up to $10\text{ }\mu\text{g m}^{-3}$ across IGP, eastern, and central India, while the $\gamma\text{HO}_2 = 0.1$ simulation shows more modest reductions in

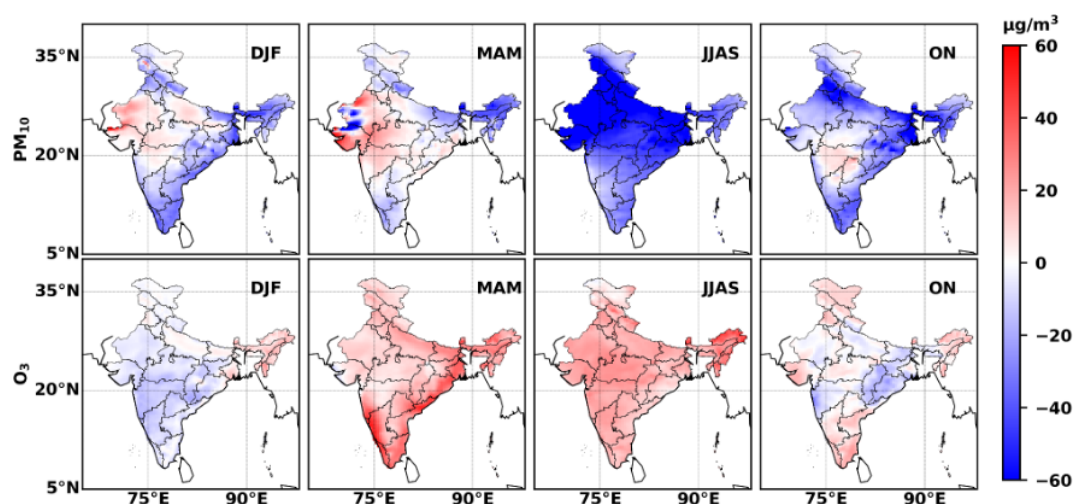


Figure 3. The difference in (top) PM_{10} and (bottom) Ozone in 2022 when compared to that of 2018. Here, DJF is December–January–February, MAM is March–April–May, JJAS is June–July–August–September and ON is October–November.

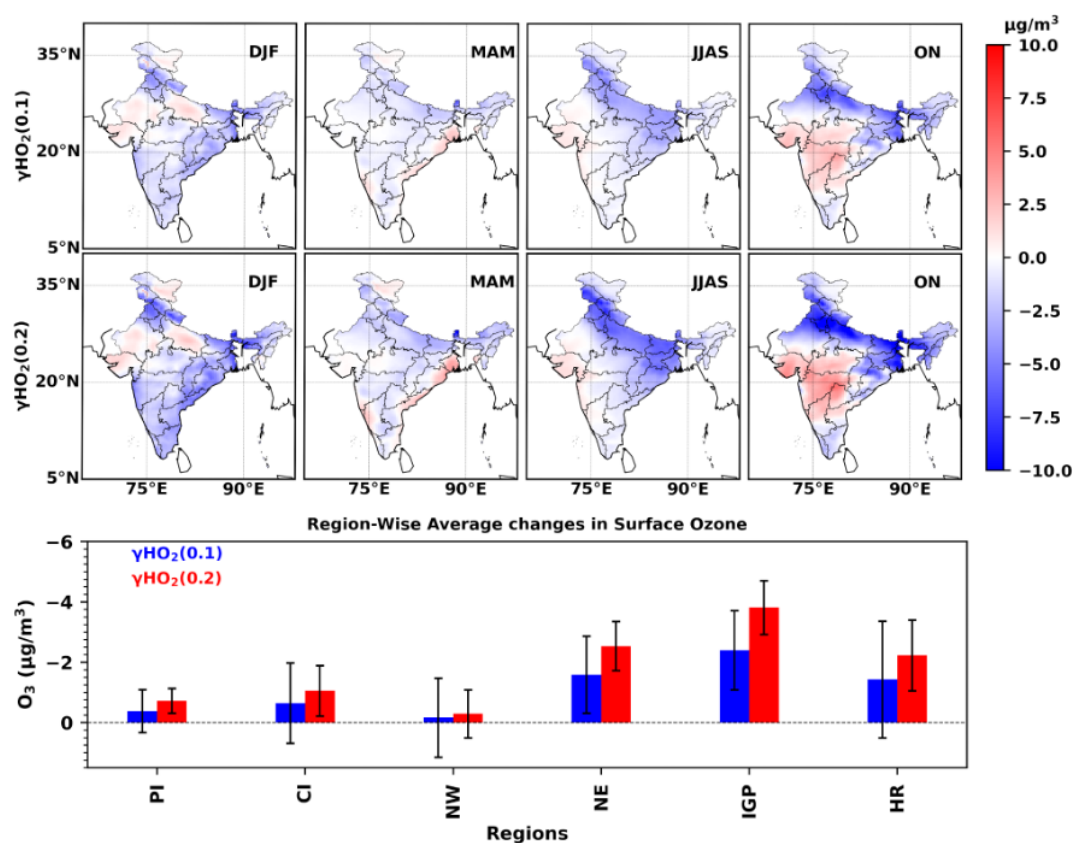


Figure 4. Difference in surface ozone response to aerosol uptake between 2022 and 2018, expressed relative to the no-uptake case ($\gamma\text{HO}_2 = 0$). The top panel shows $[(\text{O}_3 \text{ at } \gamma\text{HO}_2 = 0.0 \text{ in } 2022 - \text{O}_3 \text{ at } \gamma\text{HO}_2 = 0.1 \text{ in } 2022) - (\text{O}_3 \text{ at } \gamma\text{HO}_2 = 0.0 \text{ in } 2018 - \text{O}_3 \text{ at } \gamma\text{HO}_2 = 0.1 \text{ in } 2018)]$, while the middle panel shows the analogous calculation for $\gamma\text{HO}_2 = 0.2$. The bottom panel shows the change in surface ozone averaged for the entire year when $\gamma\text{HO}_2 = 0.1$ and $\gamma\text{HO}_2 = 0.2$, derived from the top and middle panels. The bottom panel presents a bar chart of regionally averaged annual surface ozone changes for $\gamma\text{HO}_2 = 0.1$ and $\gamma\text{HO}_2 = 0.2$. Here, negative values indicate stronger ozone suppression in 2018, while positive values indicate stronger suppression in 2022. Here, DJF is December–January–February, MAM is March–April–May, JJAS is June–July–August–September, ON is October–November, NW is Northwest, NE is Northeast India and IGP is Indo Gangetic Plains.

the range of $5\text{--}10\ \mu\text{g m}^{-3}$ over similar regions. In the MAM and JJAS seasons, the differences in ozone concentrations are comparatively weaker in both scenarios, generally within $0\text{--}5\ \mu\text{g m}^{-3}$, although localized reductions persist over northern and eastern parts of India. The interannual differences (2022–2018) shows a consistent spatial pattern, particularly when γHO_2 is 0.1, where negative anomalies of up to $7.5\ \mu\text{g m}^{-3}$ are observed across northern and central India. Additionally, small positive anomalies (up to $3\ \mu\text{g m}^{-3}$) are observed in isolated regions such as parts of western Rajasthan, Gujarat and southern peninsular India.

4.2 Surface Ozone Formation Regimes

Figure 5 shows the surface distribution of the relative chemical fluxes associated with the termination steps of ozone formation in India for the years 2018 and 2022. During monsoon in 2022, peroxy-radical self-reactions (green) dominate most of peninsular, north east and north west arid regions, while in 2018, ozone termination was equally influenced by $\text{NO}_2 + \text{OH}$ reactions and peroxy-radical self-reactions in PI, whereas north west and north east India by HO_2 uptake onto aerosols (blue). When emissions are updated to 2022 levels but meteorology is fixed at 2018 (middle column), the chemical regimes remain almost similar, indicating that meteorology contributes substantially to the changing chemical regimes, as found in the 2022 meteorology and chemistry coupled simulations (third column). The peroxy-radical self-reactions are more widespread across major cities in India (for e.g., Visakhapatnam, Delhi and Hyderabad), reflecting the reduced availability of NO_x as a radical sink. This weakening of $\text{NO}_2 + \text{OH}$ termination is particularly evident outside IGP, where declining NO_x emissions shift the balance toward radical–radical termination pathways. Thus, the dominance of peroxy-radical self-reactions indicates that NO_x emissions in these cities limit ozone formation, promoting pathways where peroxy radicals neutralize ozone, through the VOC limited ozone formation regime (Romer et al., 2018).

During winter, $\text{NO}_2 + \text{OH}$ termination and the uptake of HO_2 on aerosols became prominent over much of India, particularly in IGP. The increased aerosol surface area during this period enhances the uptake of HO_2 , thereby inhibiting ozone production (Figs. S8 and S9). The presence of aerosols, often exacerbated by biomass burning and dust transport, amplifies this effect in areas where aerosols are abundant (Ivatt et al., 2022; Singh et al., 2018). Under fixed 2018 meteorology with 2022 emissions, the chemical termination mechanisms remain similar in most regions. Nevertheless, we observe the dominance of HO_2 uptake chemical termination in the peninsular region in 2022 when model simulations are coupled with the 2022 meteorology and emissions. In post-monsoon, the termination of ozone formation shifts predominantly to HO_2 uptake across many regions, particularly in IGP, Central, and parts of peninsular India. The north east regions, however, is an exception, where

peroxy-radical self-reactions continue to dominate the termination process. This variation in termination pathways can be attributed to regional differences in aerosol loading and the influence of seasonal emissions, including biomass burning during the post-monsoon season, which contributes to higher aerosol concentrations in IGP and central I. The fixed meteorology simulations show that the spatial extent of chemical termination mechanisms closely resembles the results of the simulations with 2022 emissions and 2018 meteorology in ON.

During the pre-monsoon period in 2022, the southern parts of northern IGP are predominantly influenced by HO_2 uptake, whereas the rest of IGP and much of India find dominant termination by peroxy-radical self-reactions. The increased aerosol surface area and the presence of fire events, particularly in the dry regions of north west and north east regions, enhance the uptake of HO_2 , suppressing ozone formation (Kumari et al., 2025). Fire events and desert dust also contribute to higher aerosol concentrations, reinforcing the impact of heterogeneous reactions on ozone chemistry (Pio et al., 2008; Badarinath et al., 2010; Vadrevu et al., 2012; Aher et al., 2014; Dhanurkar et al., 2024). With 2018 and 2022 emissions under 2018 meteorology, the chemical termination in MAM is clearly through peroxy-radical reactions and $\text{NO}_2 + \text{OH}$ regimes, except in IGP where HO_2 uptake dominated the chemical termination mechanism, which can be attributed to high fire activity and dust loading (Kuttippurath et al., 2022).

Furthermore, Fig. 6 splits the domain based on the local largest termination step. In regions characterized as “ NO_x -limited,” the peroxy self-reactions dominate, making reductions in nitrogen oxides (NO_x) the most effective strategy for mitigating ozone pollution; this contrasts with “VOC limited” regimes, where VOC reductions are more beneficial, with the transition between these regimes dictated by the atmospheric concentrations of NO_x and VOCs. Aerosol uptake of hydroperoxy radicals (HO_2) on aerosol surfaces can lead to an “aerosol inhibited” environment, diminishing ozone formation in areas with elevated aerosol concentrations.

In 2018, much of IGP and central India were in an AIR during summer and monsoon, where the presence of aerosols, particularly in regions with high anthropogenic emissions and biomass burning, suppressed ozone formation by promoting the heterogeneous uptake of HO_2 . However, a substantial reduction in aerosol surface area in 2022, shifted some of these regions into a NO_x -limited regime (for e.g. eastern CI, eastern IGP), where the availability of NO_x becomes the limiting factor for ozone formation. This transition to a NO_x -limited regime is associated with a subsequent increase in ozone concentrations in 2022 compared to 2018, as the reduction in aerosols resulted in more ozone to form via the photochemical pathways involving NO_x and VOCs. The decrease in aerosol surface area, coupled with reduced aerosol-induced HO_2 removal, enhanced the ozone production efficiency, leading to higher ozone levels in these re-

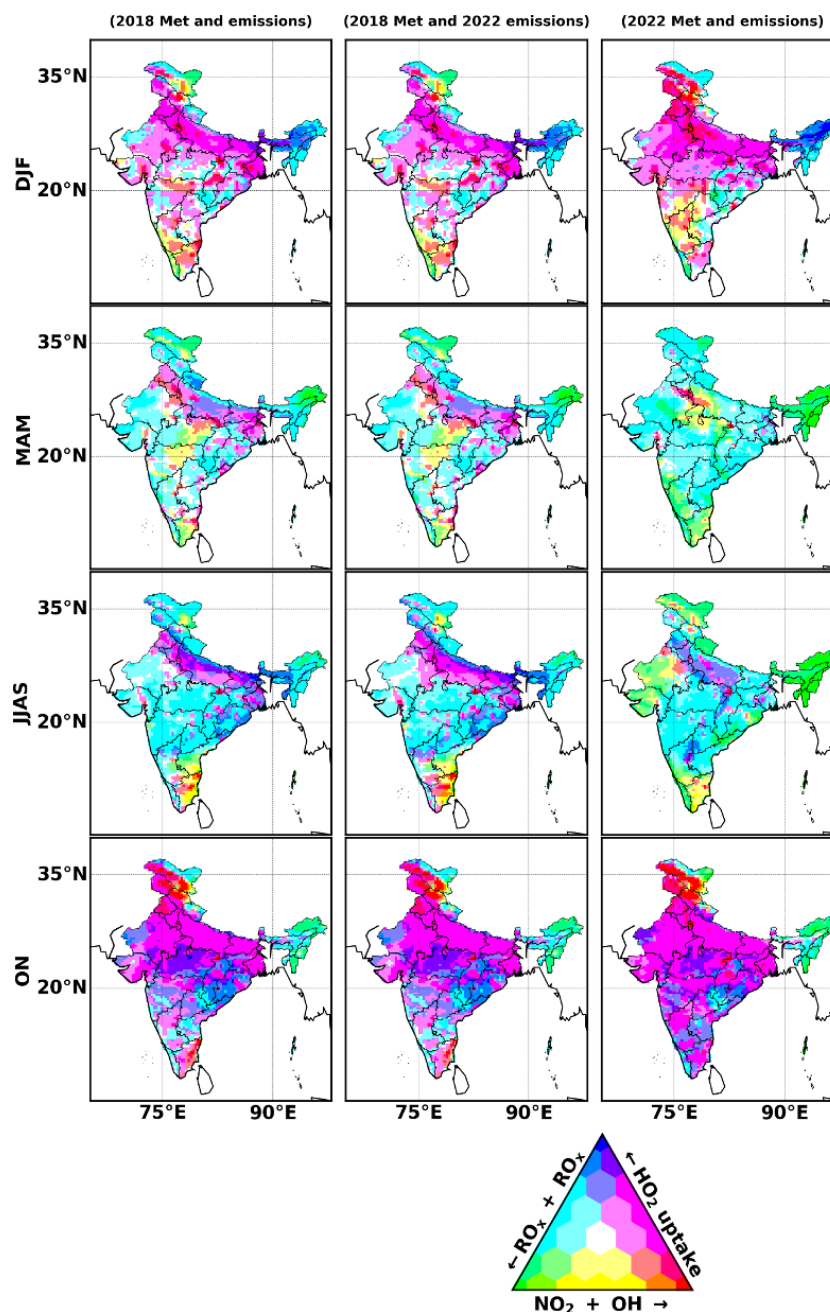


Figure 5. Mean fraction of radical termination at the surface that occurs through $\text{OH} + \text{NO}_2$ (red), peroxy-radical self-reactions (green) and aerosol uptake of HO_2 (blue) for the year (left) 2018 and (right) 2022. Here, DJF is December–January–February, MAM is March–April–May, JJAS is June–July–August–September and ON is October–November.

gions. Nevertheless, when emissions are updated to 2022 but meteorology is fixed for 2018, large parts of central, peninsular, and north east India transit out of AIR into NO_x -limited regimes. This shows that emission reductions, particularly in NO_x , are sufficient to drive regime shifts, but the meteorological changes also play a major role in these regions, with IGP remaining the strongest VOC-limited hotspot among the urban centers.

In contrast, during winter and post-monsoon, there is less to no difference in the dominant regimes between 2022 and 2018. In winter, most of India remains in a VOC-limited regime, where VOCs (such as those from biogenic sources and fossil fuel emissions) limit ozone formation. Central and north-east India continues to be in AIR during this period, with high aerosol concentrations suppressing ozone production through HO_2 uptake. During post-monsoon, the major-

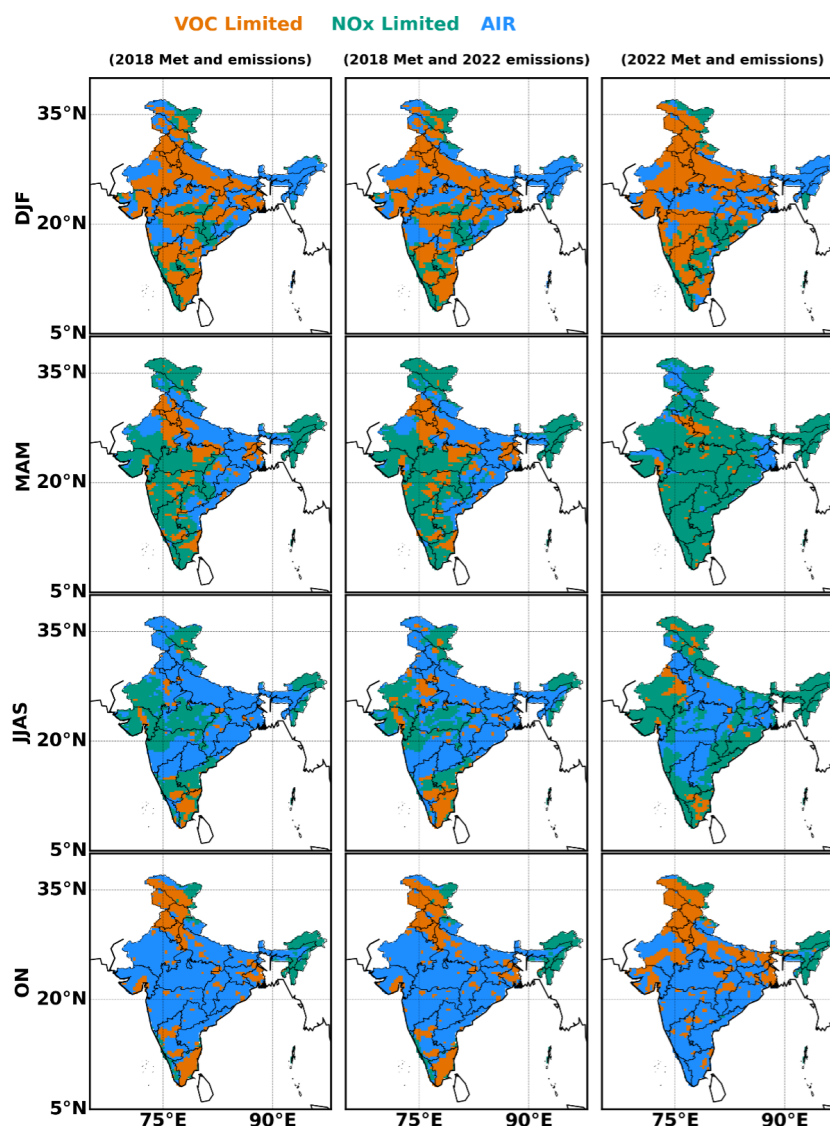


Figure 6. The regional distribution of various ozone generation photochemical regimes modelled with γHO_2 of 0.2 in (left) 2018 and (right) 2022. Here, AIR is Aerosol Inhibited Regions. Here, DJF is December–January–February, MAM is March–April–May, JJAS is June–July–August–September and ON is October–November.

ity of India remains in AIR, except for northern IGP, which is in a VOC-limited regime, and north east India, where the region is predominantly NO_x -limited. The fixed meteorology simulations show that, during the post-monsoon season, the western coast turns out to be in an AIR, whereas this region is NO_x -limited in the 2022 meteorology simulations. However, IGP remains strongly VOC-limited due to persistent high NO_x emissions and stagnant meteorological conditions, mostly during DJF and partly in ON. The seasonal consistency in winter and post-monsoon suggests that the aerosol levels together with the overall seasonal and regional variations in emissions, meteorology and chemistry play a larger role in initiating the ozone formation regime in these periods. Therefore, while aerosol changes in 2022

had a notable impact on ozone production during monsoon, they had a lesser influence on the seasonal regimes in winter and post-monsoon. This enhanced sensitivity during the monsoon is attributable to elevated humidity and aerosol liquid water content, stronger photochemical radical production, and deeper boundary layer dynamics, which collectively amplify the influence of aerosol uptake on ozone formation relative to winter or post-monsoon.

4.3 Impact of meteorology on surface ozone formation regimes

The interannual comparison shows that surface ozone variability in India is governed primarily by meteorologi-

cal drivers, with a secondary contribution from emission-induced changes in PM and associated HO₂ uptake. The difference between 2018 and 2022 (Fig. 3) shows large and spatially coherent ozone anomalies, often exceeding ± 20 – $60 \mu\text{g m}^{-3}$ across northern and central India. These differences are most pronounced during MAM and JJAS, when enhanced solar radiation, elevated temperatures, and changes in circulation patterns strongly modulate precursor availability and photochemical activity (Tyagi et al., 2020; Yadav et al., 2023; Rathore et al., 2023). For instance, ozone increases of about 30 – $40 \mu\text{g m}^{-3}$ are evident across northern India during JJAS, coinciding with large PM reductions ($> 40 \mu\text{g m}^{-3}$), while the signal is weaker in southern India (about 10 – $15 \mu\text{g m}^{-3}$). This can be attributed to the combined influence of regional circulation (more ventilated towards the southern regions versus stagnant atmospheric conditions towards the northern regions) and boundary layer dynamics (Gopikrishnan and Kuttippurath, 2024). Figures S10 and S13 show that, during the monsoon season, elevated Planetary Boundary Layer Heights (PBLH), frequently exceeding 1.0 – 1.25 km, together with relatively strong wind speeds (often greater than 4 m s^{-1} over central and southern India), enhance both vertical mixing and horizontal ventilation. In contrast, during winter and post-monsoon, a marked reduction in PBLH (generally below 0.75 km), coupled with weaker winds (commonly within the stagnant $< 3.2 \text{ m s}^{-1}$ conditions), limits atmospheric dispersion, promoting the accumulation of pollutants near the surface and thereby contributing to the pronounced regional differences in air quality. Such large magnitudes strongly suggest that the direct impact of meteorology on surface ozone pathways, through direct modulation of precursor lifetimes, photolysis rates and vertical mixing also exert dominant influence on ozone variability.

In contrast, the isolated influence of aerosol HO₂ uptake, as represented in the γHO_2 sensitivity simulations (Fig. 4), yields substantially smaller ozone changes, typically limited to about 5 – $10 \mu\text{g m}^{-3}$. This is most visible in IGP and eastern coasts of India, where high PM levels enhance HO₂ uptake, thereby suppressing ozone formation under polluted conditions. However, even in these regions, the HO₂ uptake effect is an order of magnitude smaller than the direct meteorological impact (20 – $40 \mu\text{g m}^{-3}$). The radical termination budget (Fig. 5) provides mechanistic clarity: across most of India, ozone termination is dominated by OH + NO₂ reactions (accounting for about 60% – 70%), whereas the contribution of HO₂ uptake is rarely about 20% – 30% . The HO₂ uptake pathway becomes substantial only in persistently polluted environments, such as urban regions, where high pollutant concentrations are maintained over extended periods, rather than occurring only during short episodic events (for instance, concentrations exceeding the annual average limit of $60 \mu\text{g m}^{-3}$ for PM₁₀ and the 8 h maximum of $100 \mu\text{g m}^{-3}$ for O₃). This means that while aerosol uptake of HO₂ can influence local ozone production regimes by shifting the balance between propagation and termination, its influence on

interannual ozone variability is minor compared to the large-scale and seasonally varying imprint of meteorology.

However, in future, the relative balance between these pathways is expected to evolve. As PM concentrations decline due to emission controls, the buffering capacity of aerosol HO₂ uptake will weaken, leaving ozone formation increasingly sensitive to direct meteorological perturbations. Under present conditions, the PM–HO₂ uptake pathway contributes only about 10% – 20% of the ozone variability seen in Fig. 5. However, in a future low-PM environment, even normal fluctuations in rainfall and atmospheric circulation could lead to disproportionately higher ozone. For example, during dry years with reduced wet scavenging and enhanced precursor accumulation, the absence of strong PM sinks may amplify ozone increases relative to normal or wet years, particularly in NO_x-rich urban regions of northern India. This suggests that while emissions have already reduced the potential for PM-driven radical termination, the dominant role of meteorology in setting ozone levels will likely strengthen, leading to more frequent and intense ozone enhancements during adverse meteorological episodes. Therefore, to counteract the enhanced ozone production expected under declining PM conditions, additional NO_x controls will be essential.

4.4 Impact of the national efforts in reducing NO_x and its effect on surface ozone pollution

National and local efforts to reduce NO_x emissions in India have had a noticeable impact on surface ozone pollution, particularly in urban and industrialized regions (Chen et al., 2021; Misra et al., 2021; Gopikrishnan et al., 2022). NO_x, primarily emitted by vehicular traffic, power plants and industrial processes, plays a critical role in the formation of surface ozone through photochemical reactions. By reducing NO_x emissions, several regions in India have observed a significant decrease in ozone formation. National initiatives such as stricter emission standards for vehicles, the promotion of cleaner fuels, and the adoption of advanced technologies in industrial operations have contributed to the reduction in NO_x emissions (UNESCAP, 2022). These efforts are particularly impactful in regions like IGP, which has been a hotspot for high NO_x concentrations and consequent ozone pollution. Furthermore, policies aimed at reducing vehicular traffic, such as improved public transport systems and the promotion of electric vehicles, have shown promise in lowering NO_x levels and mitigating surface ozone formation (Saikia, 2025). These measures help disrupt the photochemical cycle that leads to high ozone levels, as lower NO_x concentrations slow down ozone generation. However, the success of these efforts is often offset by increasing emissions from other sectors, such as agriculture, where practices like biomass burning continue to release NO_x (Usmani et al., 2020). Despite these challenges, sustained efforts to reduce NO_x emissions have contributed to a gradual improvement in air quality, particularly in urban centers, leading to a sec-

ondary reduction in surface ozone pollution in time. Nevertheless, more stringent emission reductions are required to tackle the increasing ozone problem in India. Therefore, we scaled down the emissions of NO_x in the model simulations to see how this decrease in NO_x would facilitate the ozone generation or destruction in different regions of India. Figure S14 shows the impact of 25 % and 50 % NO_x emission cuts on aerosol surface area. Eastern India and the IGP region experience the most substantial reduction in aerosol surface area due to NO_x reductions, especially during the post-monsoon (ON) and winter (DJF) periods. In the 50 % reduction scenario, the decrease in these areas reaches values as low as $-4 \times 10^{-7} \text{ cm}^2 \text{ cm}^{-3}$.

Figure 7 shows the difference in surface ozone in 2022 with 25 % and 50 % reduction in NO_x emissions when compared to that of 2022 with no reduction in emissions. A decrease of approximately $5\text{--}10 \mu\text{g m}^{-3}$ in ozone levels is found with a 25 % reduction in NO_x emissions, while a $10\text{--}15 \mu\text{g m}^{-3}$ decrease occurs with a 50 % reduction in NO_x emissions in India. The most substantial reduction occurs during the monsoon season, characterised by a NO_x -limited regime prevalent throughout much of India, succeeded by the post-monsoon and pre-monsoon periods. During winter, particularly in IGP and eastern central India, ozone levels increase, as these regions are limited by VOCs during this period. The increase is about $5\text{--}10 \mu\text{g m}^{-3}$ with a 25 % decrease in NO_x emissions, and about $10\text{--}15 \mu\text{g m}^{-3}$ with a 50 % reduction. Additionally, regional variations in ozone reduction are more pronounced in the 50 % decrease scenario, while the 25 % reduction shows a more uniform decrease across India ($5\text{--}10 \mu\text{g m}^{-3}$). The reduction of ozone is most noticeable in peninsular India during pre-monsoon and post monsoon ($10\text{--}15 \mu\text{g m}^{-3}$), but, the ozone reduction occurs primarily during monsoon in IGP ($10\text{--}15 \mu\text{g m}^{-3}$), with the smallest levels ($5\text{--}10 \mu\text{g m}^{-3}$) in peninsular India. During monsoon, enhanced cloud cover and aerosol–cloud interactions substantially reduce photolysis rates, while increased precipitation and convective mixing shorten NO_x and VOC lifetimes and promote vertical dilution, particularly in IGP (Liao et al., 1999; Li et al., 2016; Ojha et al., 2022). In contrast, pre-monsoon and post-monsoon conditions in peninsular India are characterised by higher solar insolation, deeper boundary layers, and weaker wet scavenging, allowing meteorology-driven changes in aerosol loading and chemical regimes to exert a stronger influence on surface ozone (Badarinath et al., 2009; Keerthi Lakshmi et al., 2024). The seasonal change can be due to decreased NO_x levels in peninsular India during the monsoon, as rainfall significantly removes precursors in the atmosphere during this season. The findings indicate that roughly 75 %–80 % of the ozone increase attributed to the decrease in PM and aerosols in India could possibly be alleviated through enhanced regulation of NO_x emissions. Also, the findings suggest that winter strategies must focus on reducing VOCs, whereas stricter NO_x control efforts are essential in other seasons to mitigate ozone pollution.

5 Conclusions

This study examines the interactions between PM_{10} , aerosols and surface ozone concentrations in India, focusing on the changes modelled between 2018, a higher PM_{10} year, and 2022, a lower PM_{10} year. The findings highlight that the seasonal and regional variability of PM_{10} and aerosol surface area play a major role in modulating surface ozone formation. During winter and post-monsoon seasons, elevated PM_{10} and aerosols, particularly in IGP and central India, led to higher uptake of HO_2 , reducing its availability for ozone production and hence suppressing ozone levels by about 30 %–40 % when compared to the surface ozone levels in 2018 and 2022. However, during monsoon and pre-monsoon seasons, reduced aerosol surface area and PM concentrations enhance photochemical ozone formation, as the decrease in aerosol surface area allows more HO_2 to participate in ozone production. The shift from an aerosol-inhibited regime to a NO_x -limited regime in 2022, due to lower aerosol pollution compared to 2018, resulted in an increase in ozone levels in several regions, especially in the IGP contributing to an increase in its levels by about $20\text{--}30 \mu\text{g m}^{-3}$. This suggests that the balance between aerosols, NO_x and VOCs dictates ozone formation, with aerosol concentrations continuing to significantly influence ozone levels during these seasons. These results are consistent with previous observational and modelling studies over India and other polluted regions, which have reported ozone suppression under high aerosol loading due to reduced photolysis and enhanced heterogeneous radical loss, and ozone enhancement following PM reductions under emission control measures, whereas this study advances earlier work by providing a nationally coherent, seasonally resolved quantification of aerosol– HO_2 uptake and meteorological influences on surface ozone. Nevertheless, local emissions from vehicular traffic, regional transportation, industrial operations, household sources and agricultural residue combustion substantially control PM_{10} and ozone levels, highlighting the necessity for thorough, region-specific emission control measures. Furthermore, scaling down the NO_x emissions in the model simulations by 25 % and 50 % shows a spatial variability in the reduction of surface ozone, with decreases of about $5\text{--}10$ and $10\text{--}15 \mu\text{g m}^{-3}$ across India, respectively. Thus, the unintended increase in surface ozone associated with declining PM levels under changing meteorological conditions can be mitigated through additional efforts in reducing anthropogenic NO_x emissions. Uncertainties related to emission inventories, representation of heterogeneous chemistry, model resolution and interannual meteorological variability may influence the magnitude of the simulated responses, but do not affect the robustness of the identified mechanisms or the relative importance of aerosol–chemistry–meteorology interactions. Therefore, this study recommends the need for integrated air quality management strategies that consider both aerosol and precursor emissions, along with regional mete-

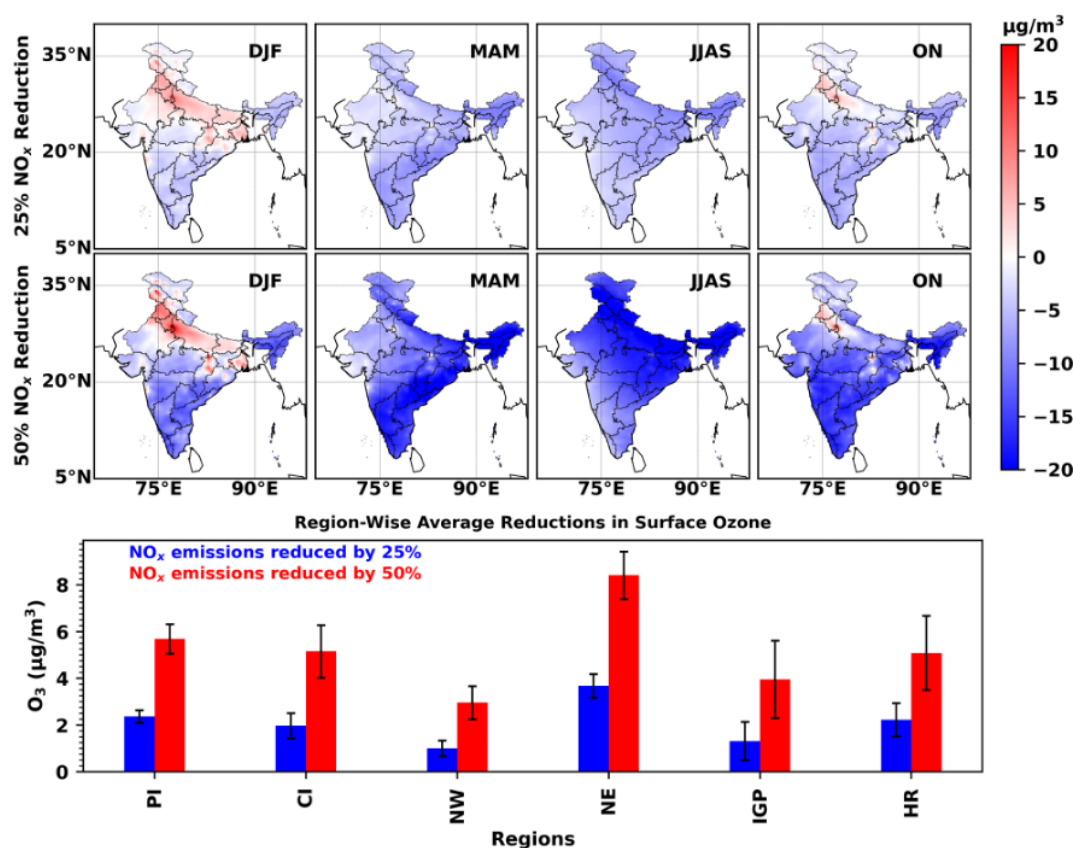


Figure 7. The change in ozone levels when NO_x emissions are scaled down by (top) 25 % and (middle) 50 % when compared to that of base simulation in 2022 with no reduction of emissions. (bottom) Region-wise averaged surface ozone reduction when NO_x emissions are scaled down by 25 % and 50 % for the year 2022. The whiskers represent the standard deviations. Here, DJF is December–January–February, MAM is March–April–May, JJAS is June–July–August–September, ON is October–November, NW is North West, NE is North East India, IGP is Indo-Gangetic Plains and HR is Hilly Regions.

orological patterns, to address ozone pollution in India effectively, highlighting that future reductions in particulate pollution, while beneficial for public health, may exacerbate surface ozone unless accompanied by coordinated NO_x and VOC mitigation under evolving climatic conditions.

Data availability. GEOS-Chem model is available via <https://geos-chem.readthedocs.io/en/stable/> (last access: 10 February 2025) and CPCB data is available via <https://app.cpcbcr.com/ccr/> (last access: 15 February 2025).

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alisation. JK: Writing – review & editing, Writing – original draft, Supervision, Visualization, Validation, Methodology, Investigation, Conceptualisation.

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