

# Seasonal Evolution of Sulfur Dioxide Over the Indian Subcontinent

*Lakhima Chutia, Narendra Ojha, Imran A. Girach, Binita Pathak, Lokesh K. Sahu, and Pradip K. Bhuyan*

**Abstract** – Reanalysis, which combines the chemical-transport model with remote-sensing measurements, has shown potential to fill data gaps over observationally sparse regions of the globe. Here, the seasonal distribution of sulfur dioxide (SO<sub>2</sub>) over the Indian region was analyzed for 2005–2015 period using Copernicus Atmosphere Monitoring Service (CAMS) reanalysis. CAMS reproduced general features and seasonality observed in surface SO<sub>2</sub> over this region. Elevated levels were revealed across the Indo-Gangetic Plain (IGP) and over eastern and central India. SO<sub>2</sub> shows a prominent seasonality over India, with a maximum typically during winter and a minimum during the summer monsoon. The winter maximum is attributed to weaker chemical sink, stagnant meteorological conditions, and elevated emissions, whereas wet scavenging, inflow of marine air, and stronger sulfate formation efficiency cause lower SO<sub>2</sub> during the monsoon. SO<sub>2</sub> levels exhibited an enhancement over IGP and central and eastern India, with rates in the range of 0.5 ppbv to 4 ppbv per decade. Like the distribution, trends also showed seasonal dependence, with weaker trends during the summer monsoon and stronger trends during the winter and postmonsoon. In situ observations of SO<sub>2</sub> and related species especially over the identified hot spots are recommended to validate satellite and model data sets and evaluate air-quality and climate implications.

## 1. Introduction

Sulfur dioxide (SO<sub>2</sub>) is an air pollutant with adverse health effects, mostly produced by the combustion of fossil fuels [1]. As a key air pollutant and aerosol precursor, SO<sub>2</sub> affects air quality, tropospheric chemistry, and the climate [2]. The chemistry of SO<sub>2</sub> in the troposphere is linked with gas-to-particle conversion processes—e.g., SO<sub>2</sub> oxidation into sulfate aerosols—as

Manuscript received 29 August 2020.

Lakhima Chutia and Pradip K. Bhuyan are with the Centre for Atmospheric Studies, Dibrugarh University, NH 37, Dibrugarh, Assam 786004, India; e-mail: chutialakhima.tsk@gmail.com, pkbhuyan@gmail.com.

Narendra Ojha and Lokesh K. Sahu are with the Space and Atmospheric Sciences Division, Physical Research Laboratory, Ahmedabad 380 009, India; e-mail: ojha@prl.res.in, lokesh@prl.res.in.

Imran A. Girach is with the Space Physics Laboratory, Vikram Sarabhai Space Centre, Veli, Thiruvananthapuram, Kerala 695021, India; e-mail: imran.girach@gmail.com.

Binita Pathak is with the Department of Physics and Centre for Atmospheric Studies, Dibrugarh University, NH 37, Dibrugarh, Assam 786004, India; e-mail: binita@dibru.ac.in.

well as formation from reduced sulfur compounds [3]. Sulfate aerosols substantially influence the earth's radiation budget by scattering solar radiation and acting as cloud condensation nuclei [4, 5]. The dominant pathway of sulfate formation is through aqueous-phase oxidation of SO<sub>2</sub> by hydrogen peroxide and ozone [3]. In the gas phase, SO<sub>2</sub> is oxidized by hydroxyl radicals (OH) to sulfur trioxide, which further leads to the formation of sulfuric acid [3].

The power-generation sector is the largest contributor (~46%) to SO<sub>2</sub> emissions in India, followed by the industrial and residential sectors [6]. Due to its relatively shorter lifetime and diverse emission sources, SO<sub>2</sub> distribution shows strong spatial heterogeneity as reported from different environments in the Indian region [7–9]. The distribution is also affected by chemistry, emissions, and seasonal changes in meteorological conditions. Due to sparse in situ measurements, an accurate understanding of seasonal changes in SO<sub>2</sub> distribution and its chemical transformations is still an intriguing challenge over the Indian region. Limited studies based on satellite observations revealed enhancements in SO<sub>2</sub> loading over the Indian region, particularly over the states of Chhattisgarh and Odisha, during the 2005–2015 period [10]. Model reanalysis, combining the chemical-transport model and remote-sensing measurements, has shown the potential to fill data gaps over observationally sparse regions of the globe [11, 12]. However, studies evaluating model reanalysis data sets against observations, seasonality in SO<sub>2</sub> levels, and its conversion into sulfate using both observations as well as model reanalysis have been still lacking over the Indian region.

Here we consider a study period of 10 years (2005–2015) to investigate seasonal variability as well as seasonal trends in SO<sub>2</sub> over the Indian region by analyzing Copernicus Atmosphere Monitoring Service (CAMS; <https://ads.atmosphere.copernicus.eu/cdsapp#!/dataset/cams-global-reanalysis-eac4-monthly?tab=form>) model reanalysis and Ozone Monitoring Instrument (OMI, [https://acdisc.gesdisc.eosdis.nasa.gov/data/Aura\\_OMI\\_Level3/OMSO2e.003/](https://acdisc.gesdisc.eosdis.nasa.gov/data/Aura_OMI_Level3/OMSO2e.003/)) satellite observations. We have estimated the efficiency of sulfate formation for different seasons.

## 2. Data and Methodology

### 2.1 CAMS Model

CAMS reanalysis provides consistent gridded fields of atmospheric composition based on an integrat-

ed atmospheric-modeling and data-assimilation system. CAMS employs the Carbon Bond 5 mechanism (CB05) chemistry module, comprising 54 chemical species and 126 chemical reactions [11, 13]. The aerosol and chemistry schemes in CB05 are independent in that both carry their own  $\text{SO}_2$  variable [14]. Anthropogenic emissions of  $\text{SO}_2$  are based on the Monitoring Atmospheric Composition and Climate/City ZEN EU projects (MACCity; <https://eccad3.sedoo.fr>) inventory available globally on a  $0.5^\circ \times 0.5^\circ$  grid. The chemical fields simulated by the CAMS model available at  $80 \text{ km} \times 80 \text{ km}$  resolution have been analyzed in this study.

## 2.2 Observational Data Sets

The present study used ground-based  $\text{SO}_2$  data reported from eight stations across the Indian subcontinent for evaluation of the model reanalysis: Nainital [15], Delhi [16], Dibrugarh [17], Kanpur [7], Patna [8], Kolkata [18], Ahmedabad [19], and Bhubaneswar [9]. Ground-based  $\text{SO}_2$  measurements have generally been performed using online analyzers based on a UV fluorescence technique.

In addition, remote-sensing based observations of the planetary boundary layer (PBL)  $\text{SO}_2$  column from the Ozone Monitoring Instrument (OMI) are analyzed here. OMI's onboard Aura is a nadir-viewing ultraviolet/visible spectrometer, which provides global coverage every day at a horizontal resolution of  $13 \text{ km} \times 24 \text{ km}$ , with a local equator crossing time at 13:45. The OMI PBL  $\text{SO}_2$  product employs an algorithm based on principal-component analysis that reduces retrieval noise by a factor of 2 [20].

## 2.3 Study Region

In this study, we divided the Indian subcontinent into five different subregions—North India (NI:  $26^\circ$ – $37.5^\circ \text{ N}$ ,  $70^\circ$ – $82^\circ \text{ E}$ ), East India (EI:  $18^\circ$ – $30^\circ \text{ N}$ ,  $82^\circ$ – $98^\circ \text{ E}$ ), Central India (CI:  $18^\circ$ – $26^\circ \text{ N}$ ,  $75^\circ$ – $82^\circ \text{ E}$ ), West India (WI:  $18^\circ$ – $26^\circ \text{ N}$ ,  $68^\circ$ – $75^\circ \text{ E}$ ), and Southern India (SI:  $8^\circ$ – $18^\circ \text{ N}$ ,  $73^\circ$ – $83^\circ \text{ E}$ )—based on the diversity of emissions, topography, and meteorology. Such classifications of different regions in the Indian subcontinent were also used in earlier studies, for example to study variations in ozone and volatile organic compounds [21, 22].

## 3. Results and Discussion

### 3.1 Model Evaluation

The performance of CAMS reanalysis was evaluated in various studies globally [11, 12]. Here we briefly describe the comparison of CAMS data with ground-based studies over the Indian region. CAMS  $\text{SO}_2$  captures observed variations to an extent, with a correlation coefficient  $> 0.55$ . Seasonal changes in  $\text{SO}_2$  have been reproduced by CAMS over the stations, including Ahmedabad, Patna, and Bhubaneswar. CAMS, in agreement with measurements, showed

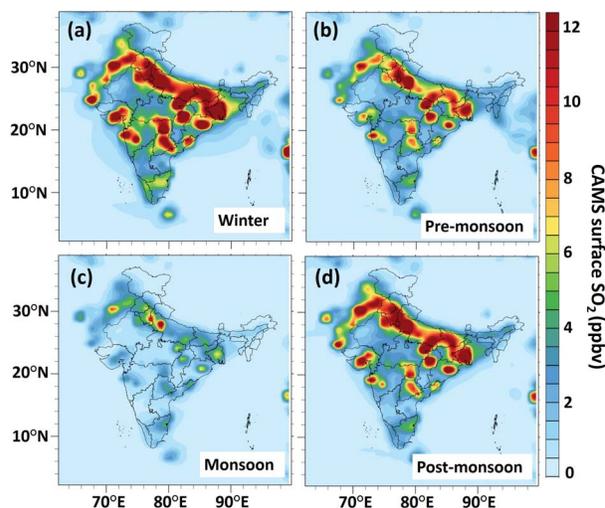


Figure 1. Mean distributions of surface  $\text{SO}_2$  from CAMS reanalysis over the Indian region during (a) winter (DJF), (b) premonsoon (MAM), (c) monsoon (JJA) and (d) postmonsoon (SON) for 2005–2015 period.

higher levels during winter/postmonsoon and lower levels during the summer monsoon. Observed  $\text{SO}_2$  values were higher in the monsoon season as compared to the winter season at complex topographical stations like Dibrugarh and Nainital, which is not reflected in the model simulation. The differences between model and observational data are suggested to be associated with uncertainties in input emissions and lack of  $\text{SO}_2$  assimilation in the model. Emission inventory data are not always consistent, particularly over rapidly developing countries such as India [23]. The lack of spatially resolved core activity data, emission factors, and removal efficiencies for different technologies create major uncertainties in emissions.

Different CAMS simulations were evaluated and discussed elsewhere [11, 13, 24, 25]. In the following sections, we analyze CAMS data to investigate the seasonal distribution and trends in  $\text{SO}_2$  over the Indian region.

### 3.2 Seasonal Distribution

Spatial distributions of surface  $\text{SO}_2$  from CAMS reanalysis during winter (December-January-February; DJF), premonsoon (March-April-May; MAM), monsoon (June-July-August; JJA), and postmonsoon (September-October-November; SON) are shown in Figure 1.  $\text{SO}_2$  levels are particularly higher over the Indo-Gangetic Plain (IGP) and the east and central parts of India compared to the southern region. Seasonally,  $\text{SO}_2$  levels are high during winter, decline during the premonsoon, and attain their lowest values during the monsoon. A recovery of higher values can be seen again during the postmonsoon. Further, these regional and seasonal differences over different subregions of the Indian subcontinent are analyzed quantitatively in section 2.3 and [21]. The month-to-month changes

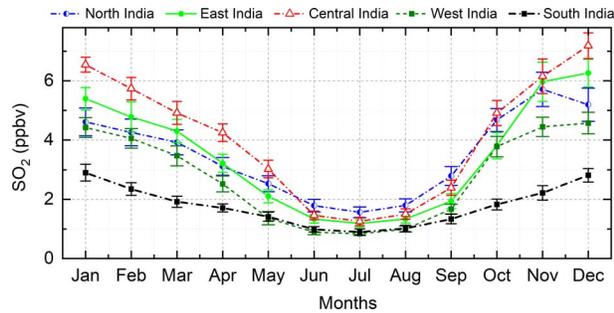


Figure 2. Monthly variation of  $\text{SO}_2$  over different subregions of the Indian subcontinent during 2005–2015 based on CAMS reanalysis.

analyzed over five different subregions are shown in Figure 2.

The highest mean  $\text{SO}_2$  mixing ratios (3 ppbv to 7.2 ppbv) can be noticed over most of the subregions during December, whereas the lowest levels are seen during July (0.8 ppbv to 1.5 ppbv). Monthly mean  $\text{SO}_2$  levels (1.2 ppbv to 7.2 ppbv) are found to be higher over Central India, owing to the enhanced energy generation and industrialization, and lower (<3 ppbv) over Southern India.

The seasonal distributions of CAMS surface  $\text{SO}_2$  and OMI PBL  $\text{SO}_2$  (Figure 3) are more or less the same, but elevated loading particularly over Chhattisgarh and Odisha, is evident in the satellite observations. Differences between reanalysis and satellite are suggested to be due to differences in the vertical layers (surface versus boundary layer) and uncertainties in retrievals and model inputs. Nevertheless, stronger  $\text{SO}_2$  levels during winter are seen in both reanalysis and satellite data. Large seasonal variability is due to the combined effect of multiple processes, such as chemistry, emissions, meteorological conditions, and boundary-layer dynamics [26]. Higher fuel usage and shallower boundary layer/stagnant condition contribute to higher levels of  $\text{SO}_2$  during winter. Another important factor that causes higher levels of  $\text{SO}_2$  during winter is the weak chemical sink, which could contribute to less-efficient  $\text{SO}_2$  oxidation. During winter, OH radicals are expected to be low due to the lower photochemical activity resulting from low incoming solar flux.  $\text{SO}_2$  levels are lowest during the summer monsoon due to efficient wet scavenging by precipitation and transport of cleaner air diluting the pollutants. Additionally, higher abundances of atmospheric oxidant (OH) can enhance the conversion of  $\text{SO}_2$  to sulfate.

To further corroborate this result, CAMS data were analyzed to estimate the sulfate formation from  $\text{SO}_2$  using the sulfate/ $\text{SO}_x$  ratio as an indicator of formation efficiency. The sulfate formation efficiency can be represented by  $\epsilon_{\text{av}} = \text{RSO}_x \times [\text{sulfate}/\text{SO}_x]$ , where  $\text{RSO}_x$  is the remaining fraction of  $\text{SO}_x$  [27].  $\text{RSO}_x$  is computed using the linear regression slope of the correlation between  $\text{SO}_x$  and CO from CAMS reanalysis and the emission ratio of  $\text{SO}_2$  to CO from the MACCcity inventory. Sulfate formation shows significant seasonal

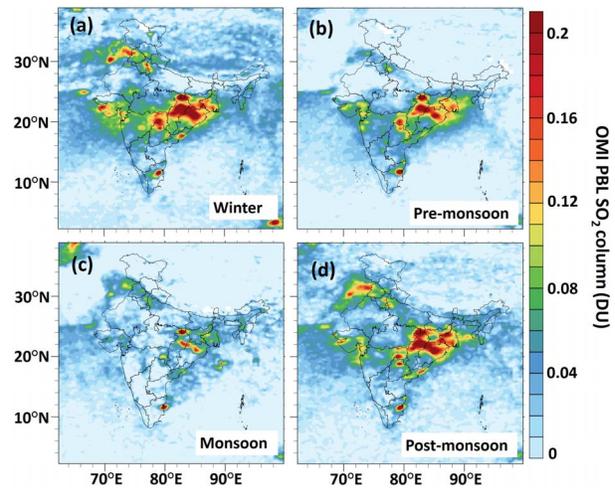


Figure 3. Mean distributions of PBL  $\text{SO}_2$  over the Indian region derived from OMI during (a) winter (DJF), (b) premonsoon (MAM), (c) monsoon (JJA) and (d) postmonsoon (SON) for 2005–2015 period.

variation, with higher efficiency during monsoon season (0.30) and lower during the winter (0.15). This implies a 30% conversion of total  $\text{SO}_2$  to sulfate during monsoon season and 15% conversion during winter. This would contribute to lower and higher values of  $\text{SO}_2$  during monsoon season and winter, respectively. Higher sulfate formation during monsoon season is attributed to stronger  $\text{SO}_2$  oxidation in the ample availability of atmospheric oxidants plus warm humid atmospheric conditions. Contrarily, during winter, limited atmospheric oxidants could reduce the formation efficiency.

### 3.3 Long-Term Trends

CAMS  $\text{SO}_2$  trends over the Indian region were examined during four different seasons (Figure 4).  $\text{SO}_2$  shows statistically significant increasing trends in the range of 0.5 ppbv to 4 ppbv per decade for all seasons over the Indian region. Increasing trends are found to be stronger over the IGP, including the industrial regions. Trends are particularly larger (1.5 ppbv to 4 ppbv per decade) during the winter and postmonsoon seasons, whereas the summer monsoon season exhibits a weaker trend (0.5 ppbv to 2 ppbv per decade) during 2005–2015, corroborating with its seasonal variation. There are some spots of decline during the winter season over some parts of West and Central India. CAMS seasonal trends are corroborated with the satellite-based trend in the PBL  $\text{SO}_2$  column (figure not shown). OMI observations show a significant increase in  $\text{SO}_2$  over East and Central India for all seasons, especially over Chhattisgarh and Odisha (> 0.2 decade/year,  $p < 0.05$ ) during 2005–2015, consistent with the model reanalysis. However, a decline is seen in the PBL  $\text{SO}_2$  column over some regions in North and West India during the premonsoon and postmonsoon seasons, which is in contrast with the model results.

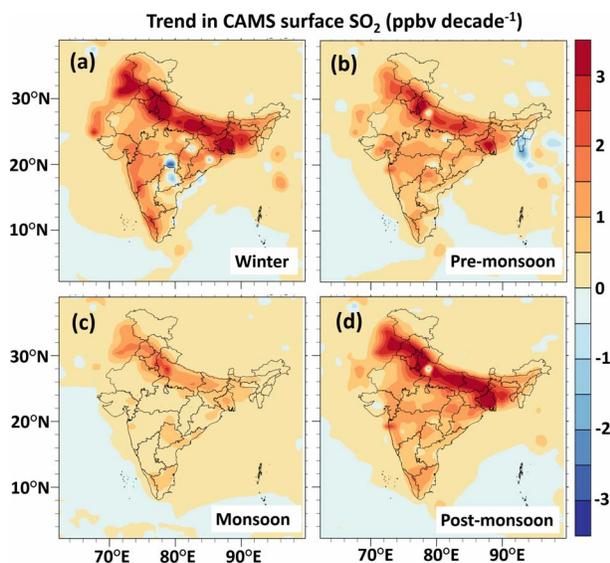


Figure 4.  $\text{SO}_2$  trends over the Indian region during (a) winter (DJF), (b) premonsoon (MAM), (c) monsoon (JJA) and (d) postmonsoon (SON) for 2005–2015 period.

Overall,  $\text{SO}_2$  enhancements are found to be in agreement with the changes in anthropogenic emissions over the study region, such as in the MACCity and CAMS inventories. This is attributed to an increase in the total installed capacity and electricity generation by coal-fired power plants during 2005–2015 over the Indian region. We additionally analyzed sulfate variation (not shown) from the CAMS reanalysis. In line with the  $\text{SO}_2$  changes, sulfate concentrations show a substantial increase with particularly strong trends of  $1 \mu\text{g}/\text{m}^3$  to  $5 \mu\text{g}/\text{m}^3$  per decade over the IGP and East and Central India. Similar to the  $\text{SO}_2$  trend, sulfate shows a stronger trend during winter and a weaker trend during monsoon season.

#### 4. Summary and Conclusions

In this study, seasonality, long-term trends, and sulfate formation efficiency of  $\text{SO}_2$  over the Indian region were investigated using model reanalysis (CAMS). The major findings of this study are the following:

- $\text{SO}_2$  showed distinct seasonal differences, with higher levels ( $\sim 3$  ppbv to 8 ppbv) during winter and lower levels ( $\sim 1$  ppbv to 2 ppbv) during monsoon season. Lower OH, stronger emissions, and stagnant meteorological conditions cause higher  $\text{SO}_2$  during winter, whereas higher sulfate formation efficiency, wet scavenging, and inflow of marine air result in lower  $\text{SO}_2$  during monsoon season.
- Significant  $\text{SO}_2$  enhancements were seen particularly over industrial and energy-generation areas during 2005–2015, with weaker trends ( $0.5$  ppbv to  $2$  ppbv per decade) during the

summer/monsoon season and stronger trends ( $1.5$  ppbv to  $4$  ppbv per decade) during the winter/postmonsoon season.

- Systematic observations together with development of up-to-date regional emission inventories are needed for assessments of air quality and climatic implications of  $\text{SO}_2$  changes over South Asia and downwind regions.

#### 5. Acknowledgments

LC is thankful to Department of Science and Technology (DST) for providing her fellowship.

#### 6. References

1. R. J. Charlson, S. E. Schwartz, J. M. Hales, R. D. Cess, J. A. Coakley Jr., et al., "Climate Forcing by Anthropogenic Aerosols," *Science*, **255**, 5043, January 1992, pp. 423-430.
2. J. T. Kiehl and B. P. Briegleb, "The Relative Role of Sulfate Aerosols and Greenhouse Gases in Climate Forcing," *Science*, **260**, 5106, April 1993, pp. 311-314.
3. J. H. Seinfeld and S. N. Pandis, *Atmospheric Chemistry and Physics: From Air Pollution to Climate Change*, 2nd Ed., New York, John Wiley & Sons, 2006.
4. S. Twomey, "The Influence of Pollution on the Shortwave Albedo of Clouds," *Journal of the Atmospheric Sciences*, **34**, 7, July 1977, pp. 1149-1152.
5. B. A. Albrecht, "Aerosols, Cloud Microphysics, and Fractional Cloudiness," *Science*, **245**, 4923, September 1989, pp. 1227-1230.
6. A. Garg, P. R. Shukla, S. Bhattacharya, and V. K. Dadhwal, "Sub-Region (District) and Sector Level  $\text{SO}_2$  and  $\text{NO}_x$  Emissions for India: Assessment of Inventories and Mitigation Flexibility," *Atmospheric Environment*, **35**, 4, 2001, pp. 703-713.
7. A. Gaur, S. N. Tripathi, V. P. Kanawade, V. Tare, and S. P. Shukla, "Four-Year Measurements of Trace Gases ( $\text{SO}_2$ ,  $\text{NO}_x$ , CO, and  $\text{O}_3$ ) at an Urban Location, Kanpur, in Northern India," *Journal of Atmospheric Chemistry*, **71**, 4, 2014, pp. 283-301.
8. S. Tiwari, P. Tunved, P. K. Hopke, A. K. Srivastava, D. S. Bisht, et al., "Observations of Ambient Trace Gas and  $\text{PM}_{10}$  Concentrations at Patna, Central Ganga Basin During 2013-2014: The Influence of Meteorological Variables on Atmospheric Pollutants," *Atmospheric Research*, **180**, November 2016, pp. 138-149.
9. C. Mallik, P. S. Mahapatra, P. Kumar, S. Panda, R. Boopathy, et al., "Influence of Regional Emissions on  $\text{SO}_2$  Concentrations Over Bhubaneswar, a Capital City in Eastern India Downwind of the Indian  $\text{SO}_2$  Hotspots," *Atmospheric Environment*, **209**, July 2019, pp. 220-232.
10. N. A. Krotkov, C. A. McLinden, C. Li, L. N. Lamsal, E. A. Celarier, et al., "Aura OMI Observations of Regional  $\text{SO}_2$  and  $\text{NO}_2$  Pollution Changes From 2005 to 2015," *Atmospheric Chemistry and Physics*, **16**, 2016, pp. 4605-4629.
11. J. Flemming, A. Benedetti, A. Inness, R. J. Engelen, L. Jones, et al., "The CAMS Interim Reanalysis of Carbon Monoxide, Ozone and Aerosol for 2003-2015," *Atmospheric Chemistry and Physics*, **17**, 3, 2017, pp. 1945-1983.
12. A. Ukhov, S. Mostamandi, N. Krotkov, J. Flemming, A. da Silva, et al., "Study of  $\text{SO}_2$  Pollution in the Middle East Using MERRA-2, CAMS Data Assimilation Prod-

- ucts, and High-Resolution WRF-Chem Simulations,” *Journal of Geophysical Research: Atmospheres*, **125**, 6 March 2020, p. e2019JD031993.
13. A. M. Inness, M. Ades, A. Agusti-Panareda, J. Barré, A. Benedictow, et al., “The CAMS Reanalysis of Atmospheric Composition,” *Atmospheric Chemistry and Physics*, **19**, 6, March 2019, pp. 3515-3556.
  14. J.-J. Morcrette, O. Boucher, L. Jones, D. Salmond, P. Bechtold, et al., “Aerosol Analysis and Forecast in the European Centre for Medium-Range Weather Forecasts Integrated Forecast System: Forward Modeling,” *Journal of Geophysical Research: Atmospheres*, **114**, D6, March 2009, p. D06206.
  15. M. Naja, C. Mallik, T. Sarangi, V. Sheel, and S. Lal, “SO<sub>2</sub> Measurements at a High Altitude Site in the Central Himalayas: Role of Regional Transport,” *Atmospheric Environment*, **99**, December 2014, pp. 392-402.
  16. A. Datta, T. Saud, A. Goel, S. Tiwari, S. K. Sharma, et al., “Variation of Ambient SO<sub>2</sub> Over Delhi,” *Journal of Atmospheric Chemistry*, **65**, 2010, pp. 127-143.
  17. B. Pathak, L. Chutia, C. Bharali, and P. K. Bhuyan, “Continental Export Efficiencies and Delineation of Sources for Trace Gases and Black Carbon in North-East India: Seasonal Variability,” *Atmospheric Environment*, **125**, B, January 2016, pp. 474-485.
  18. C. Mallik, D. Ghosh, D. Ghosh, U. Sarkar, S. Lal, et al., “Variability of SO<sub>2</sub>, CO, and Light Hydrocarbons Over a Megacity in Eastern India: Effects of Emissions and Transport,” *Environmental Science and Pollution Research*, **21**, 14, 2014, pp. 8692-8706.
  19. C. Mallik, S. Lal, and S. Venkataramani, “Trace Gases at a Semi-Arid Urban Site in Western India: Variability and Inter-Correlations,” *Journal of Atmospheric Chemistry*, **72**, 2, July 2015, pp. 143-164.
  20. C. Li, J. Joiner, N. A. Krotkov, and P. K. Bhartia, “A Fast and Sensitive New Satellite SO<sub>2</sub> Retrieval Algorithm Based on Principal Component Analysis: Application to the Ozone Monitoring Instrument,” *Geophysical Research Letters*, **40**, 23, November 2013, pp. 6314-6318.
  21. L. Chutia, N. Ojha, I. A. Girach, L. K. Sahu, L. M. A. Alvarado, et al., “Distribution of Volatile Organic Compounds Over Indian Subcontinent During Winter: WRF-Chem Simulation Versus Observations,” *Environmental Pollution*, **252**, A, September 2019, pp. 256-269.
  22. A. Sharma, N. Ojha, A. Pozzer, K. A. Mar, G. Beig, et al., “WRF-Chem Simulated Surface Ozone Over South Asia During the Pre-Monsoon: Effects of Emission Inventories and Chemical Mechanisms,” *Atmospheric Chemistry and Physics*, **17**, 2017, pp. 14393-14413.
  23. E. Saikawa, M. Trail, M. Zhong, Q. Wu, C. L. Young, et al., “Uncertainties in Emissions Estimates of Greenhouse Gases and Air Pollutants in India and Their Impacts on Regional Air Quality,” *Environmental Research Letters*, **12**, 6, May 2017, p. 065002.
  24. N. Ojha, I. Girach, K. Sharma, P. Nair, J. Singh, et al., “Surface Ozone in the Doon Valley of the Himalayan Foothills During Spring,” *Environmental Science and Pollution Research*, **26**, April 2019, pp. 19155-19170.
  25. I. A. Girach, P. R. Nair, N. Ojha, and L. K. Sahu, “Tropospheric Carbon Monoxide Over the Northern Indian Ocean During Winter: Influence of Inter-Continental Transport,” *Climate Dynamics*, **54**, 11–12, April 2020, pp. 5049-5064.
  26. N. Ojha, A. Sharma, M. Kumar, I. Girach, T. U. Ansari, et al., “On the Widespread Enhancement in Fine Particulate Matter Across the Indo-Gangetic Plain Towards Winter,” *Scientific Reports*, **10**, April 2020, p. 5862.
  27. T. Miyakawa, N. Takegawa, and Y. Kondo, “Removal of Sulfur Dioxide and Formation of Sulfate Aerosol in Tokyo,” *Journal of Geophysical Research: Atmospheres*, **112**, D13, July 2007, p. D13209.