

Influence of covariance of aerosol and meteorology on co-located precipitating and non-precipitating clouds over Indo-Gangetic Plains

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HIGHLIGHTS

- Strong aerosol-cloud relation under unstable meteorological conditions led to formation of high-level thick clouds.
- In thick clouds the activation of cloud droplets is weakly dependent on aerosols.
- Optically thin clouds led to high precipitation rate.

ABSTRACT

Aerosol-cloud-precipitation-interaction (ACPI) plays a pivotal role in the global and regional water cycle and the earth's energy budget; however, it remains highly uncertain due to the underlying different physical mechanisms. Therefore, this study aims to systematically analyze the effects of aerosols and meteorological factors on ACPI in the co-located precipitating (PCs) and non-precipitating clouds (NPCs) clouds in winter and summer seasons by employing the long-term (2001-2021) retrievals from Moderate Resolution Imaging Spectroradiometer (MODIS), Tropical Rainfall Measuring Mission (TRMM), and National Center for Environmental Prediction/National Center for Atmospheric Research (NCEP/NCAR) reanalysis-II datasets over the Indo-Gangetic Plains (IGP). The results exhibit a decadal increase in aerosol optical depth (AOD) over Lahore (5.2%), Delhi (9%), Kanpur (10.7%) and Gandhi College (22.7%) and decrease over Karachi (-1.9%) and Jaipur (-0.5%). The most stable meteorology with high values of lower tropospheric stability (LTS) is found in both seasons over Karachi. In summer season the occurrence frequency of clouds is high (74%) over Gandhi College, 60% of which are PCs. Conversely, the least number of PCs are found over Karachi. Similarly, in winter season, the frequency of cloud occurrence is low over Karachi and high over Lahore and Gandhi College. The analysis of cloud top pressure (CTP) and cloud optical thickness (COT) indicate high values of cloud fraction (CF) for thick and high-level clouds over all study areas except Karachi. The micro-physical properties such as cloud effective radius (CER) and cloud droplet number concentration (CDNC) bears high values (CER

32 $> \sim 15 \mu\text{m}$ and $\text{CDNC} > \sim 50 \text{ cm}^{-3}$) for both NPCs and PCs in summer. The AOD-CER correlation
33 is good (weak) for PCs (NPCs) in winter. Similarly, the sensitivity value of the first indirect effect
34 (FIE) is high (ranged from 0.2 ± 0.13 to 0.3 ± 0.01 in winter, and from 0.19 ± 0.03 to 0.32 ± 0.05
35 in summer) for PCs and low for NPCs. Sensitivity value for second indirect effect (SIE) is
36 relatively high (such as 0.6 ± 0.14 in winters and 0.4 ± 0.04 in summer) than FIE. Sensitivity values
37 of the aerosol-cloud interaction (ACI) are low (i.e., -0.06 ± 0.09) for PCs in summers. Furthermore,
38 the precipitation rate (PR) exhibits high values in summer season, and PR values are found high
39 in comparatively thin clouds with fewer CDNC ($< \sim 50 \text{ cm}^{-3}$) and intermediate for optically thick
40 clouds with higher CDNC ($> \sim 50 \text{ cm}^{-3}$).

41 **Keywords:** Aerosol-cloud-precipitation-interaction, Aerosol optical depth, cloud effective radius,
42 cloud droplet number concentration, lower tropospheric stability, relative humidity, first indirect
43 effect, second indirect effect, precipitation sensitivity.

44 1. Introduction

45 The aerosol-cloud-precipitation-interaction (ACPI) and aerosol-radiation-interaction (ARI)
46 significantly influence climates at the regional and global scales (Romero et al., 2021). Assessing
47 the direct and indirect effects of aerosols is crucial to understand and predict energy budget and
48 the water cycle. In the direct effect, the absorption and scattering of solar radiation by aerosols
49 lead to the warming of the atmosphere and cooling of the earth's surface (Zhou et al., 2020),
50 causing changes in the lower tropospheric stability (LTS) that further leads to modulation of
51 precipitating (PCs) and non-precipitation clouds (NPCs) (Andreae & Rosenfeld, 2008). In the
52 indirect effect, the water-soluble aerosols such as soil dust, sulfates, nitrates, and other organic
53 aerosols ejected naturally and anthropogenically serve as cloud condensation nuclei (CCN) and
54 ice nucleating particle (INP). Hence, aerosols affect the aerosol-cloud-interaction (ACI) by
55 influencing the growth of cloud droplet and cloud droplet number concentration (CDNC)
56 (Twomey et al., 1977; Albrecht, 1989; Jiang et al., 2002; Chen et al., 2011; Tao et al., 2012). The
57 increase of CDNC and decrease of cloud droplet effective radius (CER) inhibit the onset of
58 precipitation and increase the cloud lifetime (Albrecht, 1989). Conversely, the decrease in CDNC
59 and increase in CER increases the probability of precipitation rate (PR). Conversely, Stevens and
60 Feingold (2009) have shown that initially, more sea salt carried by high wind speed inhibit the

61 precipitation formation. However, the same sea spray tends to seed the coalescence by producing
62 larger CER that led to enhanced precipitation.

63 In the last few decades, most of the cultivable land of the Indo-Gangetic Plain (IGP) has been
64 replaced by urban developments. Due to the fastest growth of population, urbanization,
65 industrialization, and massive combustion of biomass and fossil fuels in residential homes and
66 factories, a decadal increase in aerosols is observed over IGP. The high aerosol loading may affect
67 the formation of tropospheric clouds and seasonal precipitation patterns (Kaskaoutis et al., 2011;
68 Singh et al., 2015; Thomas et al., 2021) and makes IGP suitable for the study of ACPI. Besides,
69 frequent variations in cloud fraction (CF), extreme precipitation and drought abrupt temperature
70 changes (e.g., heat waves), and irregular unseasonal rains may cause major and unavoidable
71 hazards at local and regional levels in the future.

72 In the last two decades, the scientific community has focused on quantification of ACI using both
73 observations (Feingold et al., 2003; Costantino et al., 2010; Zhao et al., 2018, 2020; Anwar et al.,
74 2022) and modeling techniques (Chen et al., 2016, 2018; Wang et al. 2020; Zhou et al., 2020;
75 Sharma et al., 2023). Although, a similar recent study (Anwar et al., 2022) attempted to understand
76 the sensitivities of ACI and the first indirect effect of different subsets of AOD to the different
77 conditions of RH and wind directions and found decrease (increase) in CER with aerosol loading
78 known as Twomey effect (anti-Twomey effect) over the monsoon (weak and moderately intensive
79 monsoon) regions. However, the above study excluded the other significant meteorological
80 parameters such as LTS, PR, and T_{850} and was also limited to the monsoon regions of Pakistan
81 only. Further, in the context of warm rain processes, it is generally understood that the high
82 concentration of aerosols capable of serving as CCN leads to enhanced CDNC known as the first
83 indirect effect (FIE) or Twomey effect (Twomey et al., 1977). It is also widely acknowledged that
84 CDNC plays a pivotal role in cloud microphysics and significantly influences the onset of
85 precipitation and retention of water in clouds called the second indirect effect (SIE) (Gryspeerd
86 et al., 2016; Naud et al., 2017). Whilst, in the above study, the analysis of CDNC is also not
87 addressed. Therefore, the present study aims to deepen the previous study (Anwar et al., 2022), by
88 a long-term and detailed analysis of the ACPI including aerosol-indirect effects for low-level
89 clouds extended over the whole IGP for understanding different mechanisms (condensation,
90 droplet growth and precipitation rate) of cloud and precipitation formation.

91 This study is focused on estimating the variations in sensitivities of aerosol-cloud relationship to
92 the variations in aerosol loading at specified meteorological conditions for low-level PCs and
93 NPCs in the summer and winter seasons over the IGP. This study is unique in using large number
94 of samples, classification of liquid clouds in PCs and NPCs, further classification of clouds in low,
95 mid, and high-level clouds through joint COT-CTP histograms, quantification of the sensitivities
96 of FIE, SIE, total indirect effect (TIE), and ACI to CDNC. The significant meteorological
97 parameters considered include temperature at 850 hPa, LTS, relative humidity (RH%) at 850 hPa,
98 vertical velocity (Ω), and PR. Furthermore, by utilizing the Moderate Resolution Imaging
99 Spectroradiometer (MODIS) and Tropical Rainfall Measuring Mission (TRMM) data, the
100 correlation of cloud microphysical properties (CER and CDNC) and AOD at specified values of
101 LTS and cloud liquid water path (CLWP) is examined, and precipitation sensitivity at constant
102 macro-physical condition is estimated.

103 **2. Study area and methodology**

104 2.1. Study area

105 The selected study area (Fig. 1) comprises the upper, middle, and eastern portions of the IGP. The
106 upper part consists of the densely populated and developed regions of the eastern part of Pakistan
107 i.e., Karachi (24.87°N, 67.03°E) and Lahore (31.54°N, 74.32°E) whereas the middle part comprises
108 the northern part of India i.e., Delhi (28.59°N, 77.22°E), Kanpur (26.51°N, 80.23°E), Jaipur
109 (26.91°N, 75.81°E), Gandhi College (25.87°N, 84.13°E), Kolkata (22.57°N, 88.36°E), Dhaka
110 (23.80°N, 90.41°E) and Patna (25.59°N, 85.13°E). The data analysis for the eastern part of IGP
111 (Kolkata, Dhaka, and Patna) is documented as supplementary materials.

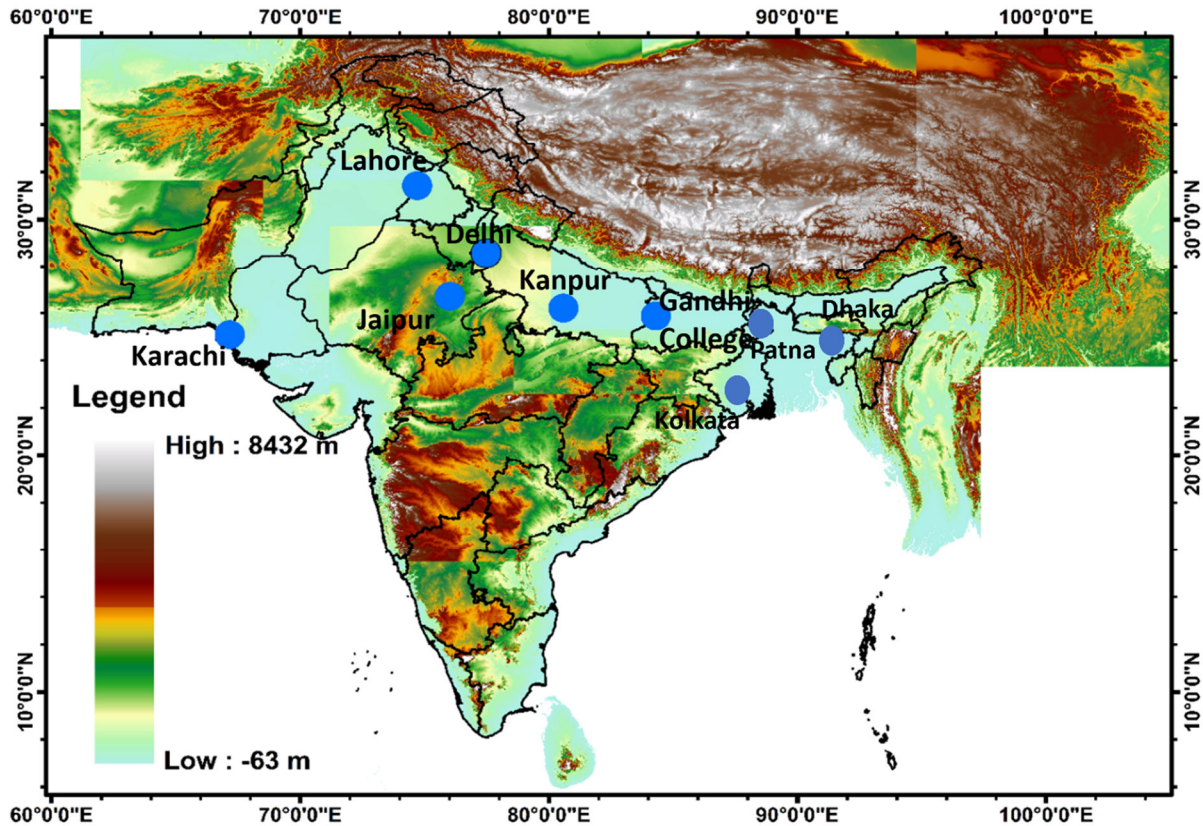


Fig. 1. Topography of the study area.

2.2. Methodology

2.2.1. MODIS, NCEP/NCAR reanalysis-II and TRMM data

Moderate Resolution Imaging Spectroradiometer (MODIS) is a major constituent of NASA's Earth Observing System (EOS). MODIS is orbiting with two onboard satellites, Terra and Aqua, launched in 1999 and 2002 respectively, with a range of 2330 km spanning the entire globe in a day. It provides data and information with a spatial resolution of 1° to study atmospheric processes and physical structure (Kedia et al., 2014; Srivastava et al., 2015). This study uses the daily mean of combined dark target and deep blue AOD at $0.55 \mu\text{m}$, cloud top pressure (CTP), cloud top temperature (CTT), CF, CER, and COT for liquid clouds from level 3 aerosol-cloud data product MOD08. Data with $\text{AOD} > 1.5$ are excluded to avoid potential misidentification of aerosols as clouds. The following adiabatic approximation (Brenquier et al., 2000; Wood, 2006; Kubar et al., 2009; Michibata et al., 2014) is used to calculate CDNC (cm^{-3}):

126
$$CDNC = \left(\frac{B}{CER}\right)^3 * \sqrt{(2 * CLWP * \gamma_{eff})}$$

127 (1)

128 Where $B = \sqrt[3]{\left(\frac{3}{4}\pi\rho_{water}\right)} = 0.0620$, ρ_{water} is the liquid water density, γ_{eff} is the adiabatic
 129 gradient of liquid water content in the moist air column (Wood, 2006). Value of γ_{eff} range from 1
 130 to 2.5×10^{-3} at a temperature of 32 K to 104 K (Brenguier, 1991; Zhu et al., 2018; Zhou et al.,
 131 2020). The CLWP is estimated by use of

132
$$CLWP = \frac{5\rho_w(CER)(COT)_w}{9},$$

133 (2)

133 Where, ρ_w is the water density at room temperature (Koike et al., 2016).

134 National Center for Environmental Prediction/National Center for Atmospheric Research
 135 (NCEP/NCAR) reanalysis datasets provide global reanalysis data sets that combine satellite
 136 observations with the simulation of models through data assimilation (Purdy et al., 2016). Daily
 137 data for meteorological parameters including temperature, RH%, and Ω at 850 hPa are retrieved
 138 at a spatial resolution of T62 Gaussian grid ($1.915^\circ \times 1.875^\circ$) from NCEP reanalysis-II datasets,
 139 and used to calculate lower tropospheric stability (LTS) defined as (Li et al., 2017):

140
$$LTS = \theta_{700} - \theta_{1000}$$

141 (3)

142 where θ is the potential temperature and the subscripts denote the pressure levels of 700
 143 hPa and 1000 hPa.

144 The Tropical Rainfall Measuring Mission (TRMM) is the first Joint satellite mission between
 145 NASA America and National Space Development Agency (NASDA) Japan, utilizing the visible
 146 infrared and microwaves to measure the rain precipitation over tropical and subtropical regions.
 147 The main TRMM instruments that are used to measure rain precipitation are precipitation radar
 148 (PR) and TRMM Microwave Imager (TMI). Where PR is operating at a frequency of 13.8 GHz
 149 and TMI is a passive microwave radiometer consisting of nine channels. A calibrated data set
 150 TRMM-2B31 of TRMM Combined Instrument (TCI) for TRMM Multi-Satellite Precipitation
 151 Analysis (TMPA) is formed from an algorithm that uses TMI and PR. The product TMPA 3B42
 152 gives the rain precipitation averages on a daily and sub-daily basis. In the current study, the data

153 product TMPA or TRMM 3B42 is used for the retrieval of PR on daily basis. The spatial resolution
 154 of TRMM 3B42 is $0.25^\circ \times 0.25^\circ$ and is available from the year 1998 to till date.

155 2.2.2. Methodology

156 The present study is designed to analyze and quantify the ACPI for PCs and NPCs in winter and
 157 summer under a variety of meteorological conditions. The daily mean data of each parameter for
 158 warm clouds are retrieved from the respective satellites and NCEP/NCAR reanalysis-II for each
 159 study site. Subsequently, statistical functions are employed to align the data collected on
 160 comparable dates. The data are then segregated into two subsets for the summer and winter
 161 seasons. Based on precipitation data from TRMM, the subsets are further divided into precipitating
 162 and non-precipitating clouds.

163 The sensitivities of cloud parameters to CDNC are analyzed through the following formulation
 164 considered from previous studies (Zhou et al., 2020):

$$165 \quad \frac{d\ln(COT)}{d\ln(CDNC)} = -\frac{d\ln(CER)}{d\ln(CDNC)} + \frac{d\ln(CLWP)}{d\ln(CDNC)}$$

166 (4)

167 In this study, the term on the left side of equation (3) is defined as total indirect aerosol effect
 168 (TIE), and the first and second terms on the right side of the equation are defined as the first indirect
 169 aerosol effect (FIE), *and second indirect effect (SIE), respectively*. Similarly, the sensitivity
 170 of CDNC to AOD is evaluated by employing the index of ACI:

$$171 \quad ACI_{CDNC} = \frac{d\ln(CDNC)}{d\ln(AOD)}$$

172 (5)

173 The sensitivity of PR to CDNC is calculated from the following equation (Jung et al., 2012) :

$$174 \quad S_0 = \left(-\frac{\partial \ln(PR)}{\partial \ln(CDNC)} \right)_{COT}$$

175 (6)

176 3. Results and Discussion

177 3.1. Regional and seasonal distribution of AOD

178 AOD is a commonly used proxy for aerosol concentration in the atmosphere and is analyzed here
179 (Fig. 2-3).

180 IGP characteristically exhibits a diverse and massive pool of aerosols due to its unique topography.
181 The western part of IGP is a coastal location and inlet for the westerly winds. Therefore, dry
182 regions and Arabian sea in the west contribute dust, sea salt and water vapors to the region. The
183 Himalayas in the north act as barriers to the winds, leading to the trapping of aerosols over the
184 central part of IGP. Therefore, this region exhibits a high concentration of anthropogenic aerosols.
185 Bay of Bengal in the east allows southeasterly winds to enter passing across Dhaka, Kolkata, Patna
186 to Delhi and Lahore (Hassan et al., 2002; Anwar et al., 2022). The westerly and easterly winds
187 traverse forested hilly terrain, rivers and lakes elevating humidity level and initiate the cloud
188 formation by activation of the newly originated small aerosol particles as CCNs and cloud
189 formation affecting the local microclimate.

190 Fig. 2 shows a decadal variation in time average maps for combined dark target and deep blue
191 AOD retrieved at $0.55 \mu\text{m}$ over the entire study area for the years (2001-2010) and (2011-2021).
192 Also, Table 1 illustrates the percentage change in decadal averaged values of AOD. The results
193 indicate that AOD exhibits a decrease over Karachi (-1.9%) and Jaipur (-0.5%). Whilst an increase
194 in AOD is observed over Lahore (5.2%), Delhi (9%), Kanpur (10.7%) and Gandhi College
195 (22.7%). Similarly, Table 1S shows the decadal change in AOD over Kolkata (18%), Dhaka
196 (22.6%) and Patna (23.3%). Similar to Gandhi College, an increase is observed over all the three
197 areas. Reason for the increase of aerosols include multiple sources of aerosols, human behavior,
198 socio-economic development at local and regional level, and unique topography for persistence
199 and retaining of aerosols.

200 Fig.3(a-b) shows the probability density function (PDF) for AOD, illustrating different
201 distributions in summer and winter season. Fig.3a shows that the distribution of AOD over Delhi,
202 Kanpur, and Gandhi College is similar. However, a shift in peak value of PDF towards high values
203 of AOD over Lahore and low values over Jaipur illustrate comparatively high and low aerosol
204 concentration in summer season over Lahore and Jaipur respectively. Likewise, Fig. 1S shows the

205 seasonal PDF values of AOD over Kolkata, Dhaka, and Patna. The results indicate similar seasonal
206 distribution functions over all the three areas of eastern IGP. In both seasons PDF peaks for high
207 values of AOD are observed over Patna showing high concentration of aerosols as compared to
208 Kolkata and Dhaka.

209 The loading of high concentration of aerosols is owing to the high-density of population,
210 industrialization, and human activities. The major sources of aerosols in all months of the year
211 include vehicular emission originated from old transport facilities, emission of smoke and soot
212 during consumption of biomass for cooking, heavy industrial emission, and aerosols produced in
213 seasonal harvesting and crop-residue burning. All these sources produce organic aerosols which are
214 characterized as hydrophilic particles and have the potential to act as CCN. Likewise, the soil dust
215 particles also act as good CCN due to their hygroscopic nature (Sun & Ariya, 2006). Moreover,
216 the meteorological conditions also play a substantial role to enhance AOD values such as the
217 uplifting of loose soil dust and swelling of aerosols due to holding the water vapors (wv) for long
218 time (Masmoudi et al., 2003; Alam et al., 2010; Alam et al., 2011;). Also, the lower but flat PDF
219 curve demonstrates low values of AOD over Karachi. Ali et al., 2020 associated the low AOD
220 values over Karachi to the westerly and southwesterly winds currents at tropospheric level.
221 However, the decreasing trend in AOD over the coastal city may also be attributed to the variations
222 in other meteorological parameters like T and RH.

223 As compared to summer season, the pattern of PDF in winter is significantly different as shown in
224 Fig. 3b. The low value of PDF (0.5) for the high value of AOD (0.9) over Karachi illustrates a
225 comparatively pristine atmosphere. Similarly, the PDF peaks for Lahore, Delhi and Jaipur (0.7, 0.7
226 and 0.8) indicate comparatively high AOD over Delhi. Likewise, the distribution over Kanpur and
227 Gandhi College is similar illustrating similar values of AOD (1.1 and 1.2 respectively). These high
228 values of AOD are attributed to the high emission of anthropogenic aerosols at local and regional
229 level over the central part of IGP (Delhi, Jaipur, Kanpur, and Gandhi College).

230 Few authors attributed the reduced values of AOD in winter season to the wet scavenging and
231 suppressed emission of aerosols from earth surface (Alam et al., 2010; Zeb et al., 2019). However,
232 in our case, the low (high) values in winter (summer) are associated to dispersion of fine (course)
233 mode particles due to the variations in meteorological conditions.

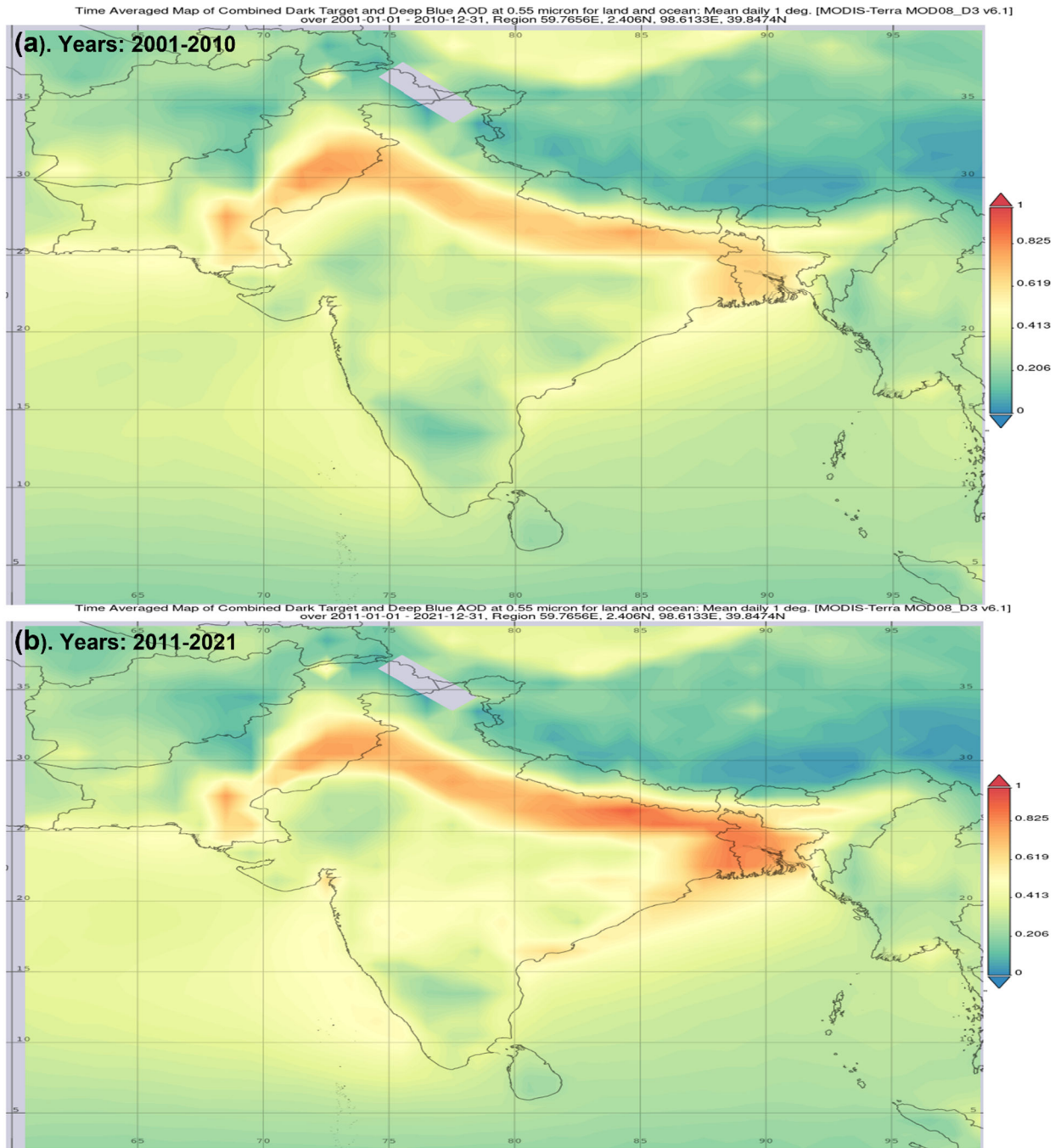


Fig. 2. Decadal increase (year: 2001-2010 and 2011-2021) in AOD over study sites.

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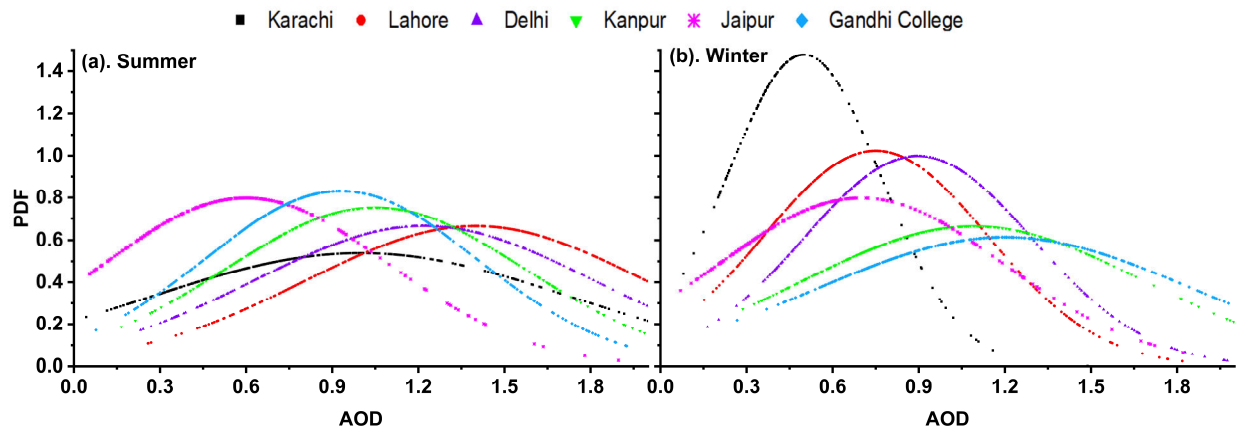
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240 **Table 1.** Decadal percentage variations in average values of AOD over all study areas

	Karachi	Lahore	Delhi	Kanpur	Jaipur	Gandhi College
Total number of counts	5902	6171	5823	5201	5907	5125
Decadal change in AOD	-1.9%	5.2%	9%	10.7%	-0.5%	22.7%

241



242

243 **Fig.3.** Probability density function (PDF) of AOD over study sites is shown (a) and (b) for summer and winter seasons
244 respectively.

245 3.2. Climatology of meteorological parameters

246 Generally, LTS has relationships to factors such as temperature, humidity, wind patterns, and
247 atmospheric pressure over extended periods. It is also widely acknowledged that atmospheric
248 stability, temperature, RH, wind speed, and direction play a significant role in cloud formation
249 (Yang et al., 2015; Tao et al., 2012). Therefore, the influence of long-term variations in the said
250 meteorological parameters are considered in the current study. The variations in meteorological
251 parameters have an unavoidable impact on ACPI. The parameters considered in this study include
252 the temperature, LTS to determine the lower atmospheric stability and instability that influence the
253 process of cloud and precipitation formation through its significant implications on evaporation
254 and convection of the air parcel, the RH% to estimate the level of wv and the Ω to assess the
255 suitable atmospheric dynamics. Fig.4 shows the variations in LTS values for NPCs and PCs in
256 winter and summer season. In winter season, the LTS values are high for NPCs and comparatively
257 lower for PCs over entire study areas. In summer season, the scenario is reversed with high values
258 for PCs but low values for NPCs, suggesting stable tropospheric layer on rainy days. This
259 stabilization may be attributed to the cold pools generated by the evaporation of falling rain

260 droplets (Wu et al., 2017). The lower LTS values for NPCs in summer season suggest the
261 likelihood of stronger instability that causes high potential of vertical motion and development of
262 thunderstorm. However, Karachi exhibits a distinct pattern of LTS with the highest values in each
263 case, which indicates the existence of the most stable tropospheric layer in Karachi due likely to
264 moist and cold sea breeze due to the city's coastal location.

265 The remaining meteorological parameters considered in this study are listed in Table 2. The high
266 values in each case are indicated in bold and low values are italicized. The results show that in
267 winter season the temperature at 850 hPa (T_{850}) is relatively high for NPCs ranging from 281 ± 2 K
268 to 286 ± 2 K. The increase in RH% for PCs during winter ranged from $(60 \pm 5)\%$ to $(72 \pm 5)\%$. Also,
269 the $\Omega > 0$ for NPCs and < 0 for PCs in winter season.

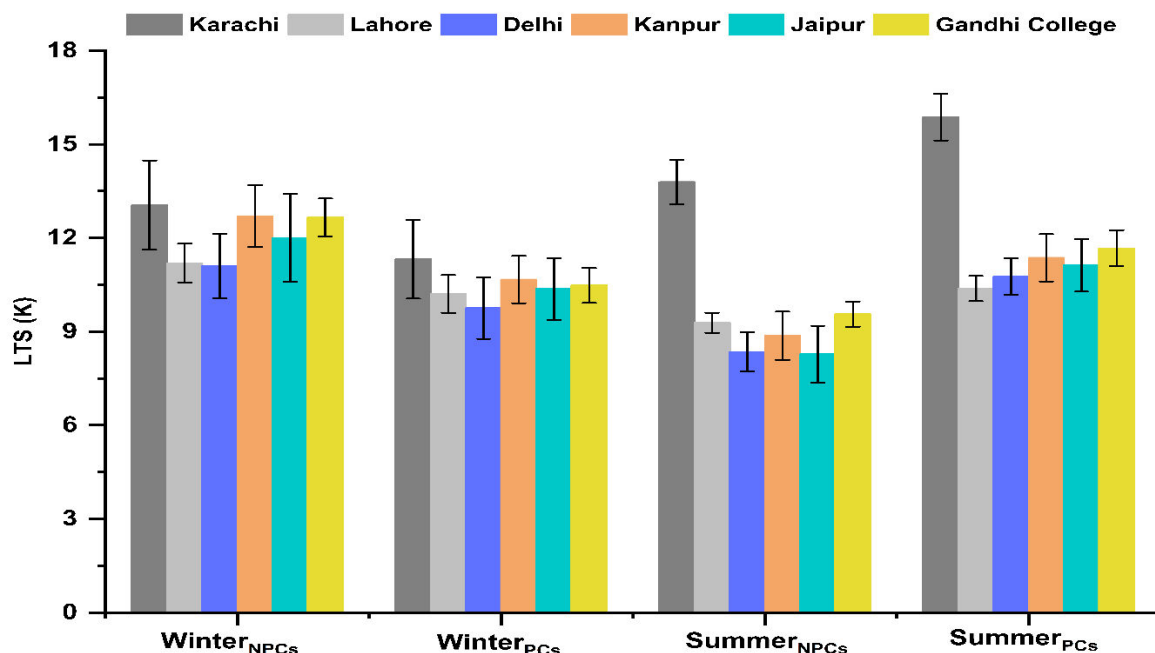
270 In summer season, it is observed that T_{850} is comparatively higher than that for the winter clouds
271 and ranged from 298 ± 0.4 to 300 ± 0.7 K and 296 ± 0.5 to 298.3 ± 0.6 K for NPCs and PCs
272 respectively. The high values of T_{850} are due to intense solar fluxes in summer season that keep the
273 temperature of the earth's surface and adjacent atmospheric layer higher. Also, the increase in
274 RH% during summer ranged between 25-45 %. The reason for the high values of wv and RH% is
275 mainly the suitable thermodynamical conditions such as evaporation and convection due to the
276 high temperature of earth surface and air (Sherwood et al., 2010). The results show high values of
277 RH% 72 ± 5 (71.6 ± 3) in winter (summer) season for PCs over Gandhi College. Conversely,
278 notable fluctuations in RH% are observed over the coastal city, Karachi, with values of 70 ± 13.9
279 (68.4 ± 6.7) in winter (summer). Similarly, Fig. 2S and Table 2S show the LTS conditions for PCs
280 and NPCs. The high LTS values indicate more stable condition over Dhaka. Similarly, Table 2S
281 shows the seasonal average values for other meteorological parameters. The results indicate high
282 values of T_{850} , RH% and Ω **297.5 ± 1.1** (297.5 ± 1.4), **82.1 ± 16.8** (76.3 ± 19.7) and **-0.2 ± 0.1** (-0.17 ± 0.1)
283 respectively for PCs (NPCs) for over Patna in summer.

284 Besides, during the last two decades, the wv and fog over the Arabian Sea were increased (Verma
285 et al., 2022). Therefore, the high values of wv and RH% in summer months is due to the high-
286 speed zonal winds that blew in the summer season and transport water vapors and sea salt from
287 the surface of the Arabian Sea and hydrophilic aerosols such as soil dust from deserts of Iran,
288 Pakistan, and India to IGP. Moreover, during the winter season, the elevated humidity levels are
289 noticeable over IGP, particularly in the vicinity of Gandhi College. This increased humidity is a

290 result of evapotranspiration driven by agricultural practices, irrigation, the presence of rivers and
 291 lakes, and the introduction of moist, cold air from western winds (Nair et al., 2020). Where $\Omega < 0$
 292 for PCs over all study areas except Karachi.

293 The distinct variations in meteorological parameters reveal the occurrence of clouds with diverse
 294 properties. The detailed analysis of such clouds is given in the next subsections.

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296

297 **Fig. 4.** Variations in lower tropospheric stability (LTS) over all study sites for PCs and NPCs in winter and summer
 298 seasons, the error bars show the standard deviation (SD) values.

299 **Table 2.** Meteorological parameters for PCs (NPCs) in summer and winter seasons. Maximum values are for both types of clouds shown in bold
 300 and minimum values are indicated as italic.

	Winter Season			Summer Season		
	T ₈₅₀ (K)	RH%	Ω (m/s)	T ₈₅₀ (K)	RH%	Ω (m/s)
Karachi	278±18 (286±2)	70±13.9 (38.3±9)	-0.04±0.04 (0.030.02)	296±0.5 (299±0.84)	68.4±6.7 (45.4±6.2)	0.02±0.04 (-5E-05±0.02)
Lahore	280.5±1 (281±2)	60±5 (36.9±4.4)	-0.04±0.04(0.07±0.02)	298.3±0.6(300±0.7)	63.6±5.2 (35±4.4)	-0.02±0.02 (0.03±0.02)
Delhi	282.5±0.9 (283±0.9)	60.3±9 (35.2±5)	-0.12±0.04(0.04±0.03)	297.4±0.6(299.3±0.8)	64.5±3 (42.5±5.5)	-0.05±0.03 (0.003±0.02)
Kanpur	284.3±0.5 (284±0.8)	64.1±4 (37±5)	-0.15±0.03 (0.04±0.04)	297±0.6(298.4±0.7)	70.4±5 (46.5±6)	-0.12±0.03 (-0.07±0.03)
Jaipur	284.4±1.3 (284±1)	67.5±6 (40.4±7)	-0.08±0.02(0.04±0.02)	296.7±0.8(299±0.9)	64.9±3 (50.6±4.2)	-0.03±0.02 (-0.03±0.01)
Gandhi College	283.3±0.5 (284.3±0.7)	72±5 (31.6±5)	-0.12±0.05(0.05±0.03)	297±0.4(298±0.4)	71.6±3 (54±4)	-0.16±0.03 (-0.01±0.03)

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3.3. Regional and seasonal distribution of clouds and precipitations

3.3.1. Regional and seasonal differences in clouds occurrence and its microphysical structure

Fig.5 shows the frequency of occurrence of precipitable clouds and total number of cloudy days. Chen et al. (2018) suggested the COT to be the effective measure for assessing the clouds and potential for precipitation. In our case, to avoid any overestimation, the COT data are aligned with PR data on corresponding dates and then filtered to include $COT \sim > 5$ for PCs. The results show that in the winter season the frequency of clouds is low over Karachi and high over Lahore and Gandhi College. The results suggest the high number of PCs only over Lahore. In summer season, the high number i.e., 74 % of the total data counts over Gandhi College are identified as cloudy days, 60 % of which are PCs. Similarly, most of the clouds over Lahore, Delhi and Jaipur are PCs. Conversely, the least number of PCs (6 %) are found over Karachi. Likewise, Fig. 3S shows the total number of cloudy days and the number of days on which PCs occurred. The high occurrence of clouds is observed over Kolkata 83% (60%) and Dhaka 91% (69%) in summer (winter) season. The high occurrence of PCs in summer is due likely to the significant impact of elevated aerosols with the southwesterly winds on the summer monsoons and occurrence of PCs. Therefore, Kolkata and Dhaka are of critical importance from perspective of aerosol loading and ACI (Dahal et al., 2022).

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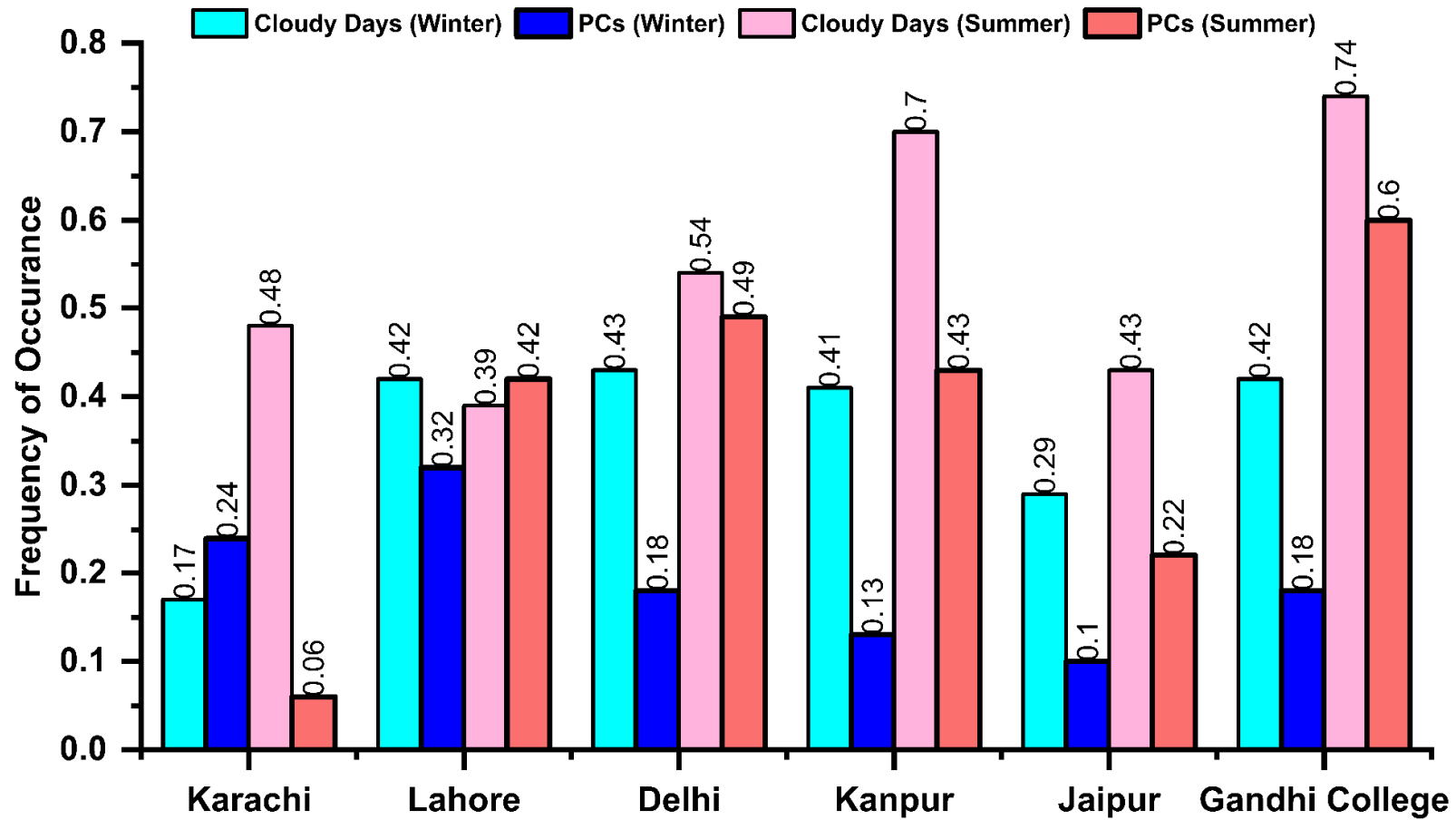


Fig. 5. Frequency of occurrence of total cloudy days (including PCs and NPCs) and only PCs is shown for both winter and summer seasons.

328 Table 3 shows the criteria adopted from previous papers (Rossow & Schiffer, 1999; Wyant et
329 al., 2006; Sharma et al., 2023) for further classification of NPCs and PCs into different type of
330 clouds. The aim of identifying the cloud types is to assess the cloud regimes and their vertical
331 structure for the better understanding of ACPI. In accordance with table 3, Fig. 6 shows joint
332 histograms of COT-CTP displaying the median values of CF for nine different types of clouds.
333 For a quick visual comparison, the cloud types are ordered from low to high level clouds. Also,
334 for each histogram, the bins of COT and CTP are located on x- and y-axis respectively. While
335 the CF of each bin is represented with the colored bar with its value mentioned in the
336 histograms as shown in Fig. 6.

337 The results exhibit noticeable differences in the pattern of cloud regimes over all study areas.
338 The diverse CF values are observed in winter and summer seasons for NPCs and PCs over
339 Karachi. In winter season, only stratus NPCs ($23 < \text{COT} < 60$, $800 > \text{CTP} > 680$ hPa) are
340 dominant with $\text{CF} \sim 0.9$. While, in summers, high value of $\text{CF} \sim 0.9$ for low and intermediate
341 thickness of high-level clouds such as Cirr-Stratus NPCs ($3.6 < \text{COT} < 23$, $180 < \text{CTP} < 440$
342 hPa) are observed. Similarly, the type of PCs in both summer and winter season that occurred
343 with $\text{CF} \sim 1.0$ include cirrus and cirr-stratus. The relatively reduced value of CF for thick NPCs
344 in winters and PCs in summers is attributed to the low values of AOD and high values of LTS.
345 The results depicted slight differences and similarities in CF values for thick and thin NPCs
346 respectively in winter season for all areas except Karachi. Besides, the high-level PCs are
347 identified in the two bins of CTP ($180 < \text{CTP} < 440$ hPa) and ($440 < \text{CTP} < 680$ hPa) over all
348 study areas. The formation of these similar types of PCs in winters are associated with the
349 similarities in Ω , LTS values and aerosols concentration.

350 Likewise, in summer season, the matrices of PCs and NPCs exhibit a wide range of cloud types.
351 However, the CF values are comparatively high for PCs. Most of the identified PCs are formed
352 in the two bins of CTP, ($180 < \text{CTP} < 440$ hPa) and ($440 < \text{CTP} < 680$ hPa) with CF values
353 ranging from 0.8 to 1.0. The results suggest low values of CF for the low-lying thick NPCs
354 over all study areas. Moreover, the results illustrate a more frequent occurrence of all the three
355 types of thick NPCs in one bin of COT ($23 < \text{COT} < 60$) and all the three types of high-level
356 NPCs for CTP ($180 < \text{CTP} < 440$ hPa) over Delhi, Kanpur, and Gandhi College. Therefore,
357 these are considered the cloudiest regimes. Besides, contrasting regional variations are also
358 observed in PCs. The maximum CF values for all types of PCs are observed over Kanpur and
359 Gandhi College. Similarly, relatively good values of CF in a bin of COT ($23 < \text{COT} < 60$) and

360 a bin of CTP ($180 < \text{CTP} < 440$ hPa) over Lahore, Delhi, and Jaipur depict the frequent
 361 occurrence of thick and high-level PCs, respectively. In addition, among all the estimated low-
 362 level PCs, cumulus and strato-cumulus exhibit good CF values (0.7) over Kanpur and Gandhi
 363 College. The formation of thick clouds can be attributed to the enhanced convection process
 364 due to the atmospheric instability.

365
 366

Table 3. Classification of clouds based on CTP – COT joint histograms.

CTP (hPa)	COT		
	0-3.6	3.6-23	23 to >60
440 to <180	Cirrus	Cirr-Stratus	Deep convection
680-440	Alto-Cumulus	Alto-Stratus	Nimbo-Stratus
<800 to 680	Cumulus	Strato-Cumulus	Stratus

367

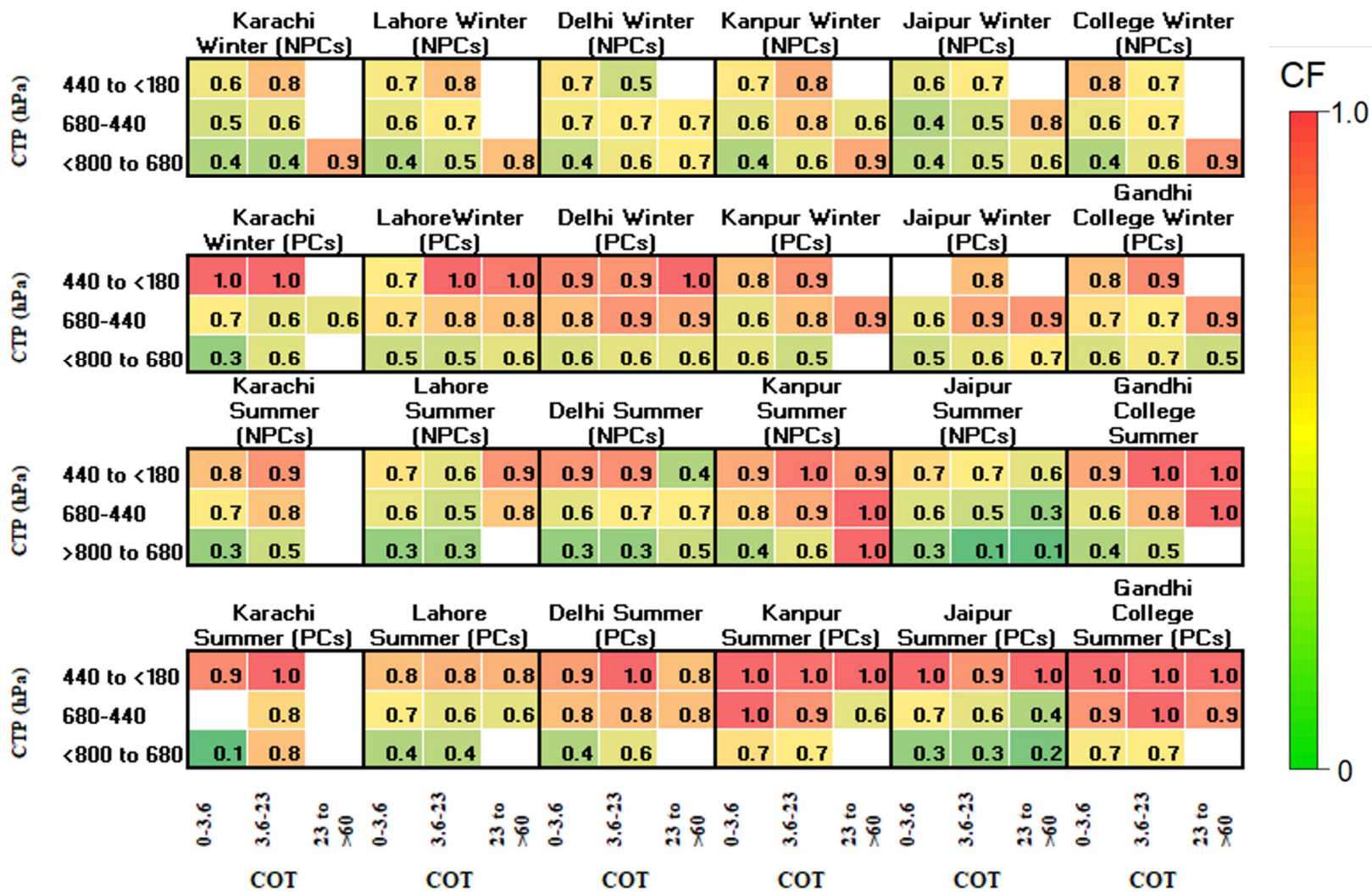


Fig. 6. Types of NPCs and PCs in winter and summer season

368

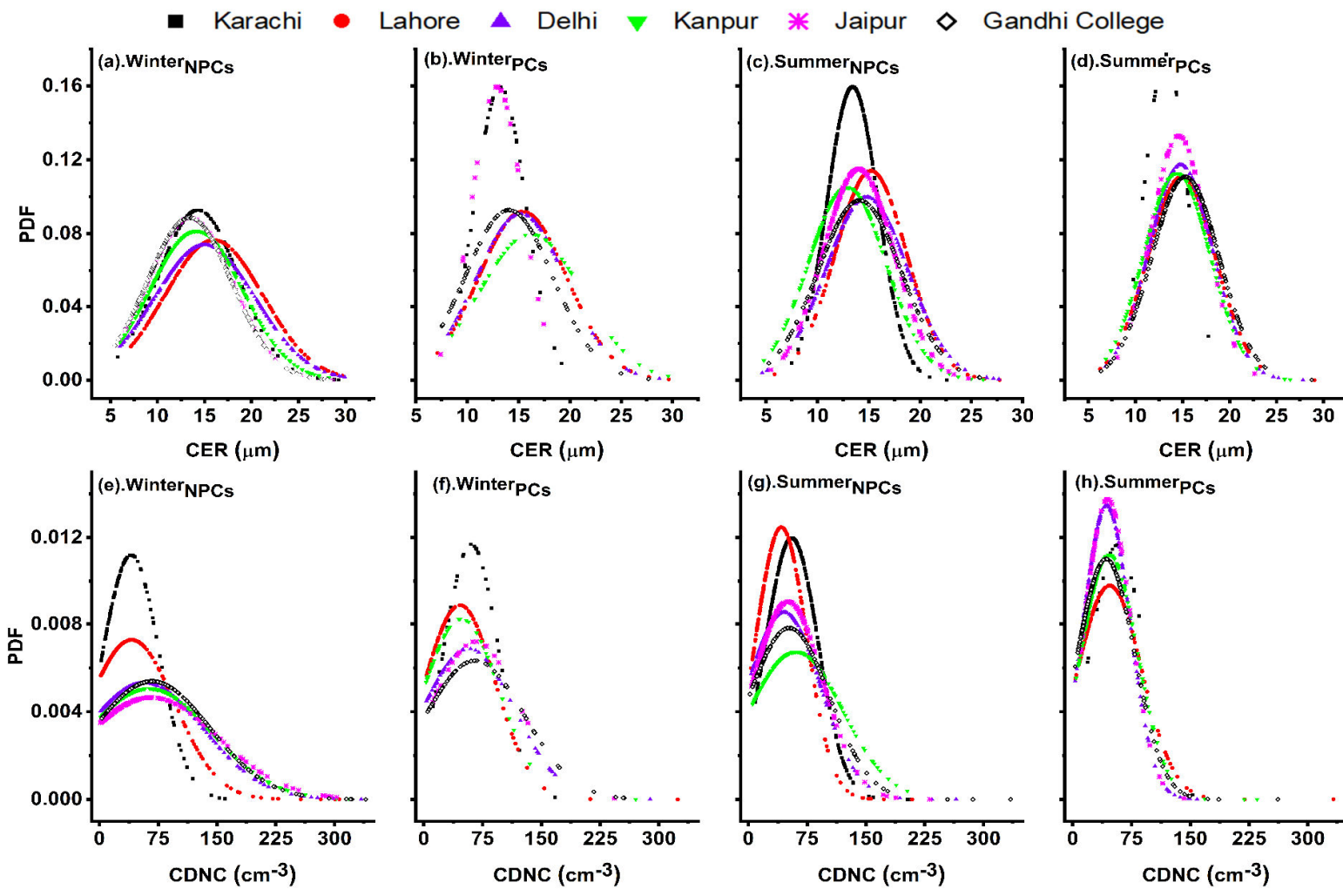
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370 After estimating the cloud types, Fig. 7 shows the probability distribution function (PDF) of
371 cloud microphysical properties for the identification of differences in microstructure of NPCs
372 and PCs in summer and winter seasons. From the results it is depicted an approximately similar
373 pattern for the CER of NPCs in winter. However, the clouds have high peaks of PDF for lower
374 values of CDNC over Karachi. The low number of CDNC results in thin NPCs as shown in
375 Fig.7. Similarly, Fig. 7(c and g) shows the microstructure of NPCs in summer. The results
376 indicate that as compared to CER values in winter, the probability of CER $> \sim 15 \mu\text{m}$ is high in
377 summer season. However, high peak for CER $< 15 \mu\text{m}$ is observed over Karachi. Similarly, the
378 CDNC shows a high probability for CDNC $> 50 \text{ cm}^{-3}$ with the high PDF values over Karachi.
379 Where, the lowest number of CDNC is observed over Lahore indicating the formation of high-
380 level thin NPCs in summer.

381 Fig. 7(b and f) shows the distribution pattern of CER and CDNC of PCs in winter season. It is
382 clearly observed that the distribution of CER for PCs is like those for NPCs in winter season.
383 However, PDF have peak values for relatively higher CDNC, which illustrates the occurrence
384 of thick clouds. Fig. 7(d and h) shows the variations in CER and CDNC in summer season. The
385 results show a wider distribution for CER $> \sim 15 \mu\text{m}$ and higher peaks for CDNC $> \sim 50 \text{ cm}^{-3}$
386 suggesting the formation of thick PCs in summer as shown in Fig.6.

387

388



389

390

Fig. 7. Probability density function (PDF) of precipitating (PCs) and non-precipitating clouds (NPCs) in winter and summer season

391 3.4. Aerosol-Cloud-Precipitation Interaction (ACPI)

392 In the following sections, ACPI is analyzed and discussed in detail for PCs and NPCs in
393 summer and winter seasons.

394 3.4.1. Aerosol effects on cloud properties

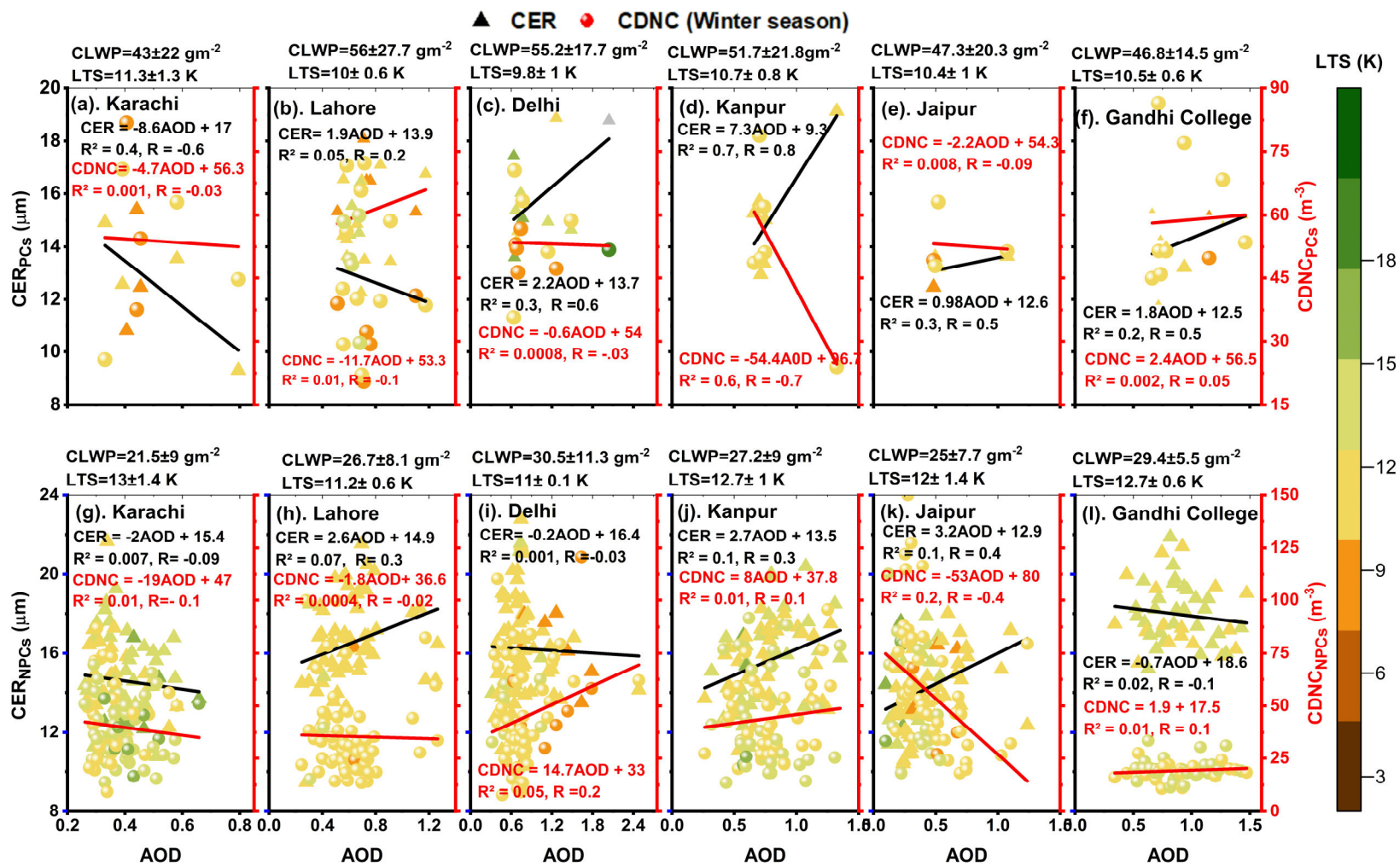
395 The impact of aerosols on CDNC and CER of PCs and NPCs is illustrated as scatter plots in
396 Fig. 8-9. The quantification of the AOD-CER and AOD-CDNC relationships is demonstrated
397 through detailed linear regressed slopes, regression coefficients (R^2) and Pearson's correlation
398 coefficient (R). The color bar represents the variations in LTS. Fig.8 shows that in winter
399 season, the AOD-CER correlation is good for PCs and weak for NPCs. The results also show
400 that the LTS values are higher for NPCs. The weak AOD-CER correlation may be linked to the
401 inhibition of droplet growth due to less soluble aerosols, originated from biomass burning
402 (Kang et al., 2015). In our case, all the selected study areas are among the most urbanized and
403 industrialized areas of IGP. Therefore, most of the prevailing aerosols are the less soluble soot
404 and BC particles. That weakened activation of cloud droplets inhibits the formation of PCs and
405 evaporates to higher altitudes and thereby increases the droplet residence time (Kumar &
406 Physics, 2013). Besides, the results show a contrasting pattern of LTS values. Although, RH
407 over Karachi ($38.3 \pm 9\%$) is higher than over the other study areas (shown in Table 2), the
408 negative AOD-CER correlation is observed over Karachi due to its coastal location, the low
409 value of AOD and high level of LTS.

410 Fig. 9 illustrates the AOD-CER and AOD-CDNC correlation in summer season. The results
411 depict a more significant and positive AOD-CER correlation in summer season than winter
412 season. Unlike winter season, the high LTS values are observed for PCs. Yuan(2008)
413 associated the positive AOD-CER correlation to the soluble organic aerosols. Myhre et al.
414 (2007) hypothesized that the positive AOD-CER correlation is a maximum for low CTP and
415 minimum for high CTP. Hence, in our study, referable to the approximated CF values shown
416 in Fig.6, the significant and positive AOD-CER correlation under unstable atmospheric
417 conditions resulted in thick and high-level clouds. Furthermore, it is observed that CER and
418 CDNC values for NPCs increase with increasing instability. Meanwhile, the enhanced process
419 of droplet activation may result in large AOD, higher CER, giant and fewer CCN (Yuan, 2008).
420 Therefore, the weak correlation of AOD with CER and CDNC may be due to the
421 anthropogenically ejected water-soluble organic aerosols and a smaller number of CCN.

422

423 Fig. 5S and 6S show the impact of AOD on CER and CDNC for PCs and NPCs in winter and
424 summer respectively. The results indicate a positive and weak AOD-CER correlation 0.2, 0.07
425 and 0.004 for NPCs over Kolkata, Dhaka, and Patna respectively and for PCs (0.08) over both
426 Kolkata and Patna. Similarly, a positive and weak AOD-CDNC is observed over all areas for
427 PCs. Likewise, Fig. 6S also illustrates weak AOD-CER correlation is 0.06, 0.2 and 0.12 for
428 both types of clouds in summer. As compared to other areas, the correlation analysis is less
429 significant over Karachi, Kolkata, Dhaka and Patna. This can be attributed to the persistence
430 of diverse aerosol types influenced by their coastal locations, different meteorology and the
431 alternating inflow and outflow of easterly and westerly winds.

432 Recent advances in remote sensing led to cost-effective solutions and an increase in available
433 data at various temporal and spatial resolution to bridge scientific gaps among different
434 disciplines. While satellite-based retrievals have many advantages over in-situ and ground-
435 based measurement such as broader regional coverage and enhanced spatial resolution, they
436 are still prone to considerable uncertainties owing to the indirect nature of remote-sensing,
437 retrieval algorithms, thermal radiance, infrequency of satellite overpasses, and cloud top
438 reflectance (Hong et al., 2006; Tian et al., 2010; Hossain et al., 2006). In our study, apart from
439 the aforementioned factors contributing to the uncertainty, any residual cloud contamination
440 could also lead to biased retrieval of AOD. Likewise, satellite-based retrievals for cloud
441 properties are crucial to understanding the pivotal role of clouds in climate and the role of
442 clouds is still a dominant source of uncertainty in prediction of climate change. These,
443 uncertainties in AOD and retrievals of cloud properties also propagate through the modeling
444 process, potentially leading to less accurate climate predictions. Likewise, these uncertainties
445 appeared to influence the findings in the current investigation. For instance, a limited
446 correlation between AOD and CER is observed over Lahore, particularly in cloudier regimes
447 as depicted in Fig. (5-6). This contrasts with robust impacted documented in the earlier studies
448 (Michibata et al., 2014). However, high sensitivity of SIE is observed for PCs particularly in
449 winter season indicating the delay in onset of precipitation and more retention of clouds.



450

451

Fig. 8. AOD-CER and AOD-CDNC regression and correlation coefficient for PCs and NPCs over all study areas in winter season.

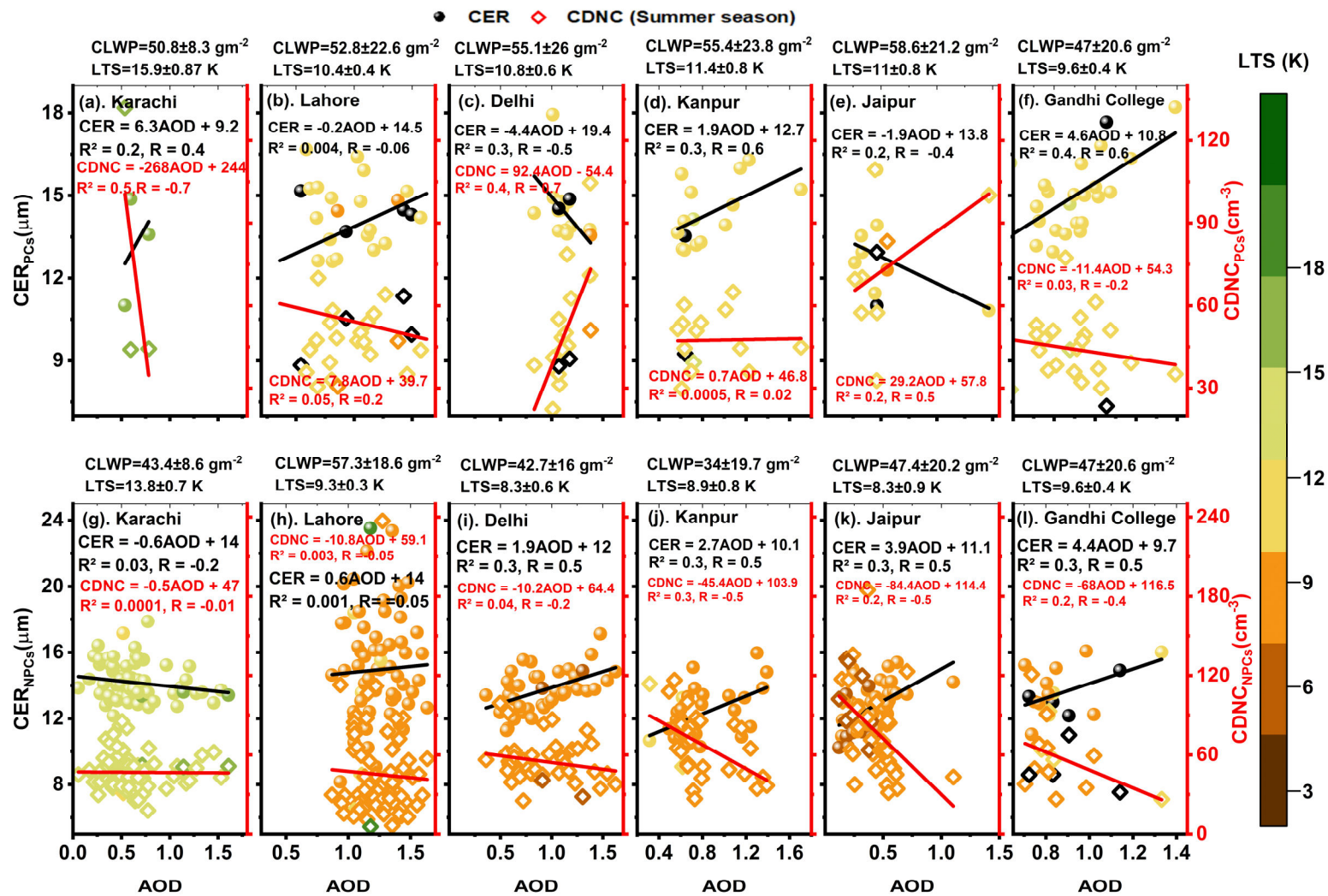


Fig. 9. Same as Fig. 8 but in summer season.

452

453

3.4.2. Seasonal variations in sensitivities of aerosol-cloud indirect effects and ACI

454
455 Fig.10 shows assessment of four ACI sensitivities in terms of CDNC using daily mean values
456 of MODIS observations available over the entire study area. Since studying the effects of
457 aerosols on the co-located clouds is a challenging task due to the overestimation of thin clouds
458 in AOD retrievals. Therefore, to minimize the propagation of AOD retrieval errors in ACI, the
459 current study attempted to estimate the sensitivities of different cloud mechanisms to CDNC.

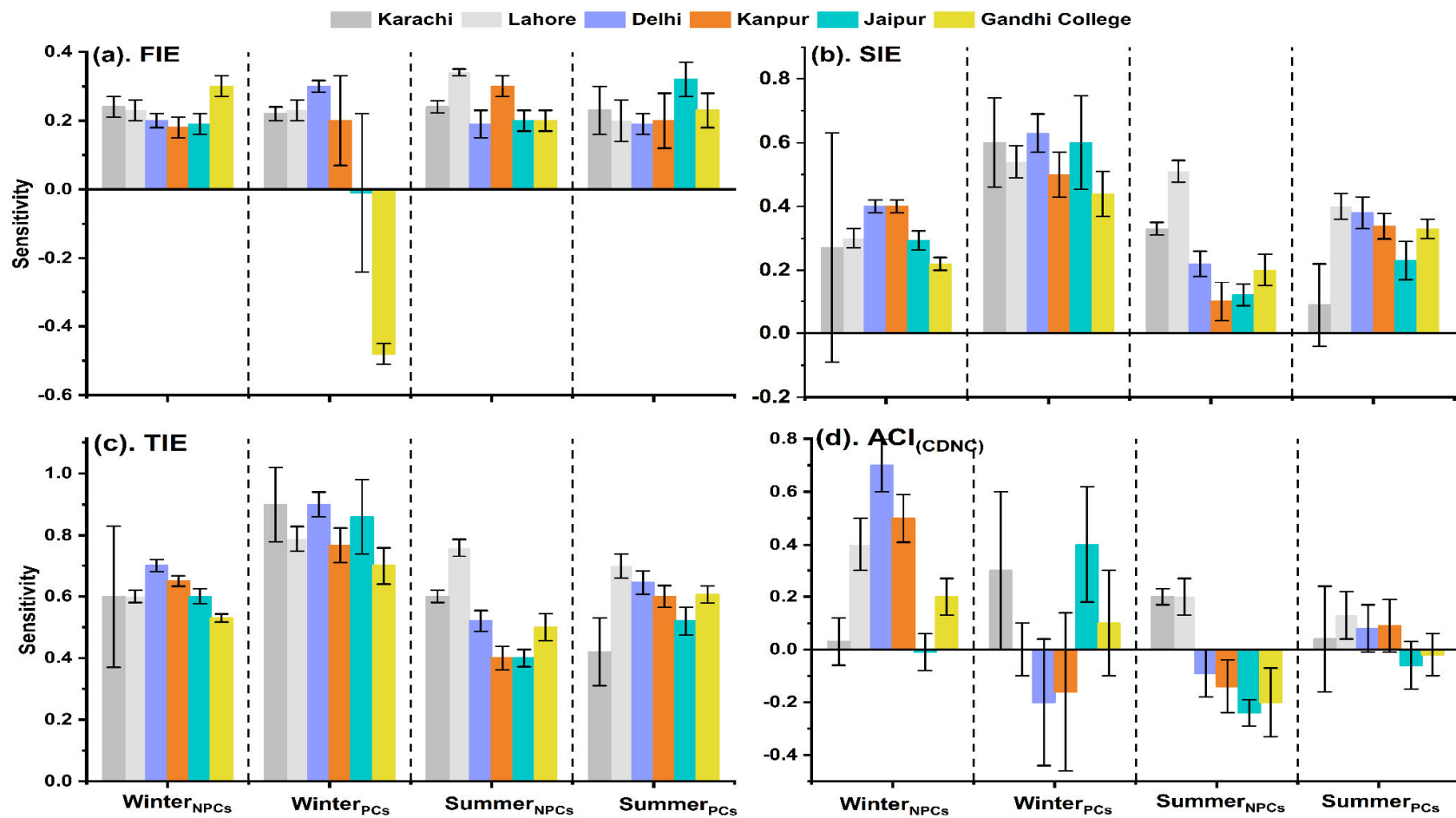
460 The sensitivity of CER to CDNC is assessed as a signature of FIE as shown in Fig.10a. The
461 positive values illustrate that CER decreases with an increase in CDNC revealing the
462 occurrence of the Twomey effect. Whilst the negative values depict the anti-Twomey effect.
463 Tripathi et al., (2007) divided IGP into four regions as western, central, eastern part of IGP and
464 the foothills of Himalayas. Their results depicted high concentration of dust in western part,
465 and an increase in anthropogenic aerosols as one moves from western to eastern part of IGP.
466 Therefore, they attributed the resulted strong indirect effect in winter to the high concentration
467 of regional anthropogenic pollution. However, in our case, the FIE is investigated for both PCs
468 and NPCs in both seasons. The resulted approximations in winter season show strong (weak)
469 sensitivity of FIE for PCs (NPCs). Similarly, the estimated sensitivity of FIE for all NPCs and
470 PCs is also positive in the summer season. Fig. 7S(a) shows sensitivities for FIE in both seasons
471 for PCs and NPCs. The results indicate high values of sensitivity FIE in winter season which
472 is similar to the results for Karachi, Lahore, Delhi and Kanpur as shown in Fig. 10 a. This is
473 attributed to high level of aerosol emission from residential heating and industrial activities.
474 Furthermore, the results illustrate higher values of FIE in summer. This is attributed to the
475 massive aerosol loading due to aerosol carried by winds and originated by anthropogenic
476 activities and unstable meteorology.

477 Fig. 10b illustrates the sensitivity of CLWP to CDNC as a proxy for evaluation of the SIE or
478 lifetime effect. The positive sensitivity estimated for all NPCs and PCs suggested that the
479 CLWP increase with increase in aerosol. Further, the results show that the sensitivity of SIE is
480 stronger for PCs in winter which indicates the delay in onset of high PR. Similarly, the results
481 show that the SIE sensitivity values are higher for PCs than for NPCs in the corresponding
482 seasons. Therefore, the results depict that the lifetime of PCs is greater than NPCs. Which is
483 attributed to the high level of RH for PCs as shown in Table 2. Fig. 10 (a and b) shows that the
484 FIE sensitivities are weaker than SIE.

485 Fig. 10c shows the TIE in terms of the sensitivity of COT to CDNC. The results illustrate
486 positive values of sensitivity for all NPCs and PCs which indicate that COT increases with
487 increase in aerosol concentration. The results also reveal that sensitivity of TIE is a linear sum
488 of the sensitivities of FIE and SIE. Further, the results also suggest that the variations in TIE
489 sensitivity are largely dependent on SIE.

490 Fig.10d shows the sensitivity of CDNC to AOD as an estimation of ACI in terms of CDNC.
491 The positive values show the increase in CDNC with the increase in AOD. Therefore, positive
492 ACI reflects the inhibition of precipitation formation. Whilst the negative values illustrate the
493 decrease in CDNC and enhanced PR (Fan et al., 2018). The results depicted relatively large
494 and positive sensitivities for NPCs in winter over Lahore, Delhi, and Kanpur, which inhibits
495 the onset of rainfall. The Sensitivity of ACI for NPCs in summer is positive over Karachi and
496 Lahore and negative over Delhi, Kanpur, Jaipur, and Gandhi College. Ackerman et al. (2004)
497 associated the negative ACI_{CDNC} to the wet scavenging and mixing of air by entrainment. In
498 our case, negative ACI may be due to the growth of CER and decrease in CDNC with aerosol
499 loading under unstable conditions (shown in Fig. 9). Further, the magnitude of sensitivity for
500 PCs in summer is low. Which can be due to the droplet growth through collision coalescence
501 and wet scavenging in thick clouds, decreased dependency on CCN.

502



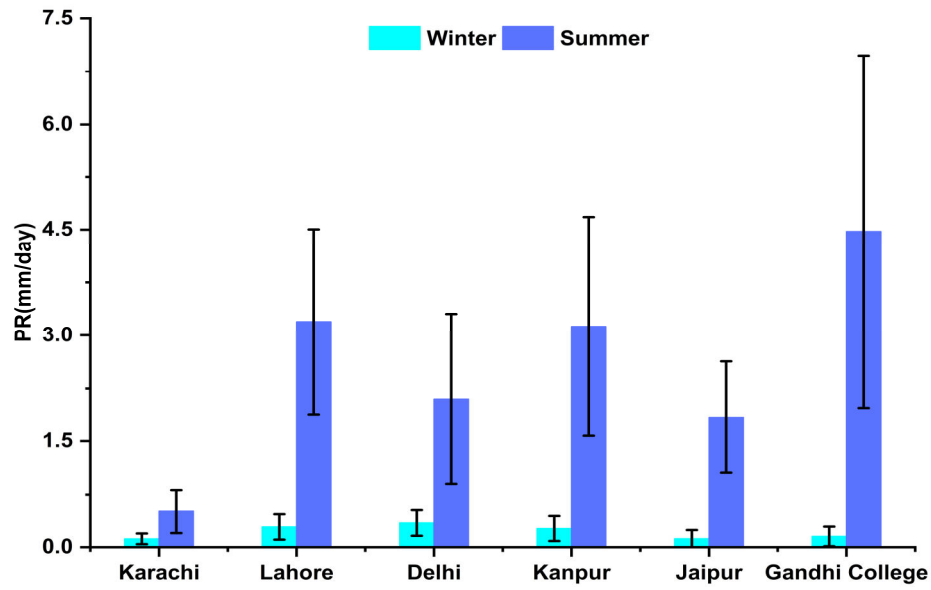
504

505 **Fig. 10.** The sensitivity metrics estimated for aerosol-cloud relationship using CDNC is shown in (a) $FIE = -\left(\frac{\partial \ln(CER)}{\partial \ln(CDNC)}\right)$ (b) $SIE = \left(\frac{\partial \ln(cLWP)}{\partial \ln(CDNC)}\right)$,
 506 (c) $TIE = \left(\frac{\partial \ln(COT)}{\partial \ln(CDNC)}\right)$ and (d) $ACI = \left(\frac{\partial \ln(CDNC)}{\partial \ln(AOD)}\right)$. Where, the error bars show the standard deviation (SD).

507 *3.4.3. Aerosol effects on precipitation*

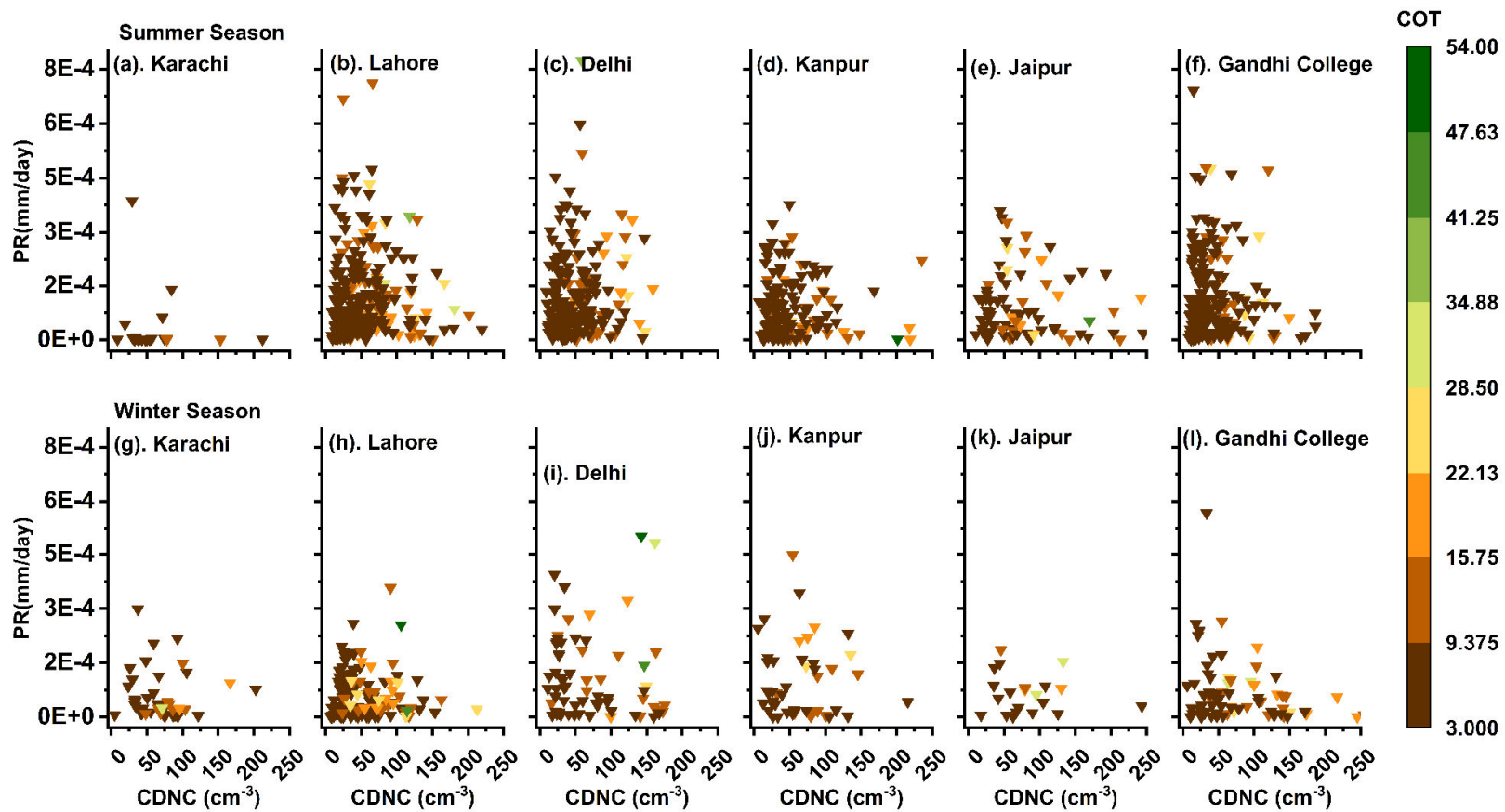
508 Fig. 11 shows the average values of PR in mm/day retrieved from TRMM. The results show
509 an obvious seasonal difference in precipitation occurrence. The reason for the high (low) PR
510 values is due to the suitable meteorological condition including high (low) LTS values for PCs
511 in summer (winter) season (shown in Fig. 8-9). The stable atmospheric condition with high
512 LTS value in winter serves to inhibit the convection process and have a significant impact on
513 controlling the PR in winter (Zhao et al., 2006). Conversely, during summer season, the
514 meteorological instability prevails with low LTS values which result in high RH. This not only
515 causes enhanced AOD due to the water uptake and resulted swelled hydrophilic aerosols (Alam
516 et al., 2010; Alam et al., 2011) but also affects the cloud and precipitation formation due to the
517 enhanced evaporation and convection. Additionally, Fig. 8-9 also show evidently and
518 specifically during summer that the possible cause of positive AOD-CER correlation is the
519 negative AOD-CDNC correlation under unstable meteorology over all areas except Karachi.
520 As a result, Fig. 11 shows high (low) values of PR over all areas with maximum over Gandhi
521 College (Karachi). The results show high (low) approximation of PR over Gandhi College
522 (Karachi). Knowing that the rate of conversion of CDNC to precipitation is proportional to
523 CER (Wolf & Toon, 2014). Therefore, the high PR values is due to the growth of bigger cloud
524 droplets in summers. Further, apart from the reasons mentioned in the preceding sections, the
525 other justification for the differently perturbed aerosols, clouds, and precipitation pattern over
526 the study areas in summers is due to the entrance of southeast winds from Bay of Bengal
527 passing across Gandhi College to Delhi and Lahore and entrance of same winds from Arabian
528 sea to Pakistan through Karachi (Anwar et al., 2022).

529 Fig. 12 shows scatter plots of PR verses CDNC. The plot is colored with COT to examine the
530 impact of CDNC on PR for similar macrophysics. When CDNC are few, then the COT are
531 sparse that grow larger, form less reflective clouds and precipitate faster (Kump & Pollard,
532 2008). The same phenomenon seems true in our case. The results illustrate high PR (0.0007
533 mm/day) values for clouds with COT ranging from 3 to 28 with $CDNC < \sim 50 \text{ cm}^{-3}$ and
534 intermediate for optically thick clouds and $CDNC > \sim 50 \text{ cm}^{-3}$ in both seasons.



535

536 **Fig. 11.** Mean Precipitation rate (PR) for the PCs in winter and summer season and SD values
 537 with 95% confidence interval.



538

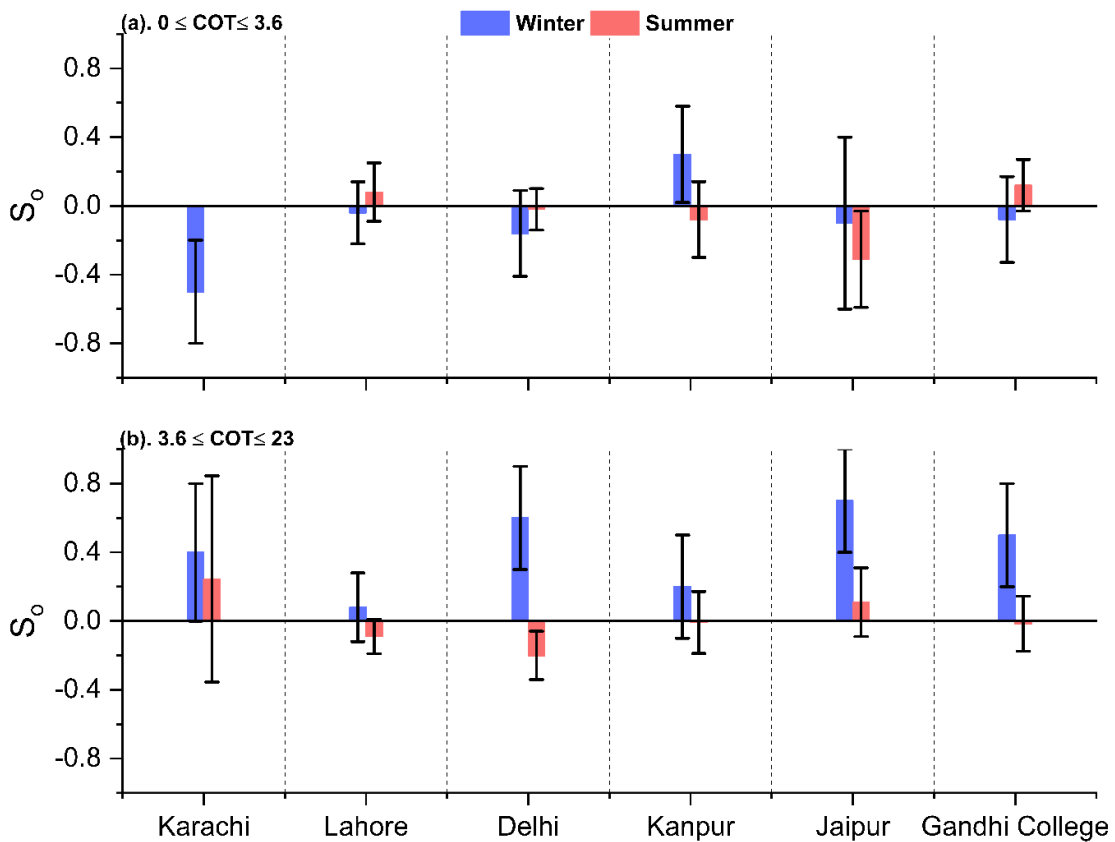
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Fig. 12. Scatter diagrams of PR (mm/day) verses CDNC (cm^{-3}) in summer and winter seasons. color coding shows the COT of PCs.

540

541 Fig.13 shows the sensitivity (S_o) of PR to CDNC defined by $S_o = \left(-\frac{dln(PR)}{dln(CDNC)}\right)_{COT}$ for clouds
 542 of low and intermediate thickness illustrated in Fig. 13 a and Fig. 13 b respectively. However,
 543 sensitivity analysis for $COT > 23$ could not be performed due to less number (0 to 04) of
 544 available samples. In the sensitivity equation the minus sign shows the suppression of
 545 precipitation formation due to the increase in CDNC. Further, when S_o is positive, correlation
 546 between PR and CDNC is negative; however, for negative S_o , PR and CDNC are positively
 547 correlated. The results show peak values of S_o i.e., 0.7 ± 0.3 , 0.6 ± 0.3 , 0.5 ± 0.3 , and $0.4 \pm$
 548 0.4 over Jaipur, Delhi, Gandhi College and Karachi respectively at intermediate values of COT
 549 in winter, indicating the occurrence of lightly precipitating clouds. Referable to Fig. 13b, the
 550 low magnitude of S_o 0.2 ± 0.3 and 0.08 ± 0.2 over Kanpur and Lahore respectively is due to
 551 coagulation, in which precipitations are less sensitive to CDNC.

552



553

554 **Fig. 13.** Sensitivity ‘ S_o ’ of precipitation rate (PR) for two bins of COT shown in (a). $0 \leq COT$
 555 ≤ 3.6 and (b). $3.6 \leq COT \leq 23$.

556

557 **4. Conclusion**

558 In this study, the long-term (2001-2021) data retrievals from MODIS coupled with TRMM and
559 NCEP/NCAR reanalysis-II datasets over the entire study area are compiled and analyzed for
560 PCs and NPCs in winter and summer season. The following are the main findings of this study.

561 A decadal decrease in AOD is observed over Karachi (-1.9%) and Jaipur (-0.5%). Meanwhile,
562 AOD exhibits an increase over Lahore (5.2%), Delhi (9%), Kanpur (10.7%) and Gandhi
563 College (22.7%). The LTS values are High (low) for NPCs (PCs) in winter and for PCs (NPCs)
564 in summer season. However, among all study areas, Karachi exhibits comparatively high LTS
565 values in both seasons. Apart from that, the increase in RH% for PCs ranged from 33-57% in
566 winter and from 25-45 % in summer. $\Omega > 0$ for all NPCs in winter and < 0 for PCs in both
567 winter and summer seasons.

568 In winter season low frequency of cloudy days over Karachi and high over Lahore and Gandhi
569 College is estimated. Also, the high number of PCs are estimated only over Lahore. In summer
570 season, out of the 74 % of the cloudy days, 60 % are PCs over Gandhi College. Similarly, most
571 of the clouds over Lahore, Delhi and Jaipur are PCs. Conversely, the least number of PCs (6
572 %) is found over Karachi. The high-level PCs are identified in one bin of CTP ($180 < \text{CTP} <$
573 440 hPa) over all study areas in winters. In summer season, all the three types of high level and
574 thick PCs have significant values of CF. The low-level PCs are identified as stratus clouds.
575 Further, PDF values for $\text{CER} > \sim 15 \mu\text{m}$ and $\text{CDNC} > \sim 50 \text{ cm}^{-3}$ for NPCs and PCs is high
576 (low) in summer (winter) over all areas except Karachi.

577 The AOD-CER correlation is good for PCs and weak for NPCs in winter season. Also, the CER
578 and CDNC values increase with increase in LTS. The sensitivity value of FIE is high (low) for
579 PCs (NPCs) in winter. Further, magnitude of sensitivity of FIE (SIE) is low (high). Also, the
580 sensitivity of TIE is a linear sum of the sensitivities of FIE and SIE. Further, ACI sensitivity
581 values for PCs in summer are small, illustrating less dependency of CER on CDNC in thick
582 clouds.

583 The high (low) PR values are observed in summer (winter). Further, high PR values for
584 comparatively thin clouds with fewer $\text{CDNC} < \sim 50 \text{ cm}^{-3}$ and intermediate for optically thick
585 clouds and $\text{CDNC} > \sim 50 \text{ cm}^{-3}$ are observed. Sensitivity values are small (high) for thick clouds
586 in summer (winter).

587 Being one of the major source regions of anthropogenic aerosols across the globe, IGP
588 offers interesting insights into the study of ACPI coupled with aerosol indirect effects. This
589 study highlights that the aerosol-cloud relationship exhibits different behavior under
590 different meteorological conditions, at coastal and inland locations. Thus, compared to
591 other study areas, the stable atmospheric conditions due to the constant sea breeze
592 weakened the ACI over Karachi, which resulted in a smaller number of CDNC, NPCs, and
593 PCs. Further, our study also provides a very good platform for the detailed analysis of
594 sensitivity tests of aerosol indirect effects and precipitation formation.

595 **Limitations and future recommendations:**

596 Although the current study is as thorough as possible, however, it has its limitations due to
597 the topographical complexity of IGP, the lack of in-situ measuring instruments in Pakistan,
598 and the intrinsic uncertainties associated with satellite-based data. Therefore, simulations
599 of ground-based measurements along with satellite-based retrievals and calculation of
600 cloud properties and CCN by different Community Atmosphere Model (CAM) and
601 Weather Research and Forecasting (WRF) Models are recommended for deep insight into
602 the various mechanisms of ACPI over IGP.

603 **Data Availability:** The MODIS and TRMM data can be obtained from the NASA Goddard
604 Earth Sciences Data and Information Center (GES DISC) and can be retrieved from the
605 websites: <https://modis.gsfc.nasa.gov/data/> and <https://gpm.nasa.gov/data> . The reanalysis-
606 II datasets are obtained from the website:
607 <https://psl.noaa.gov/data/gridded/data.ncep.reanalysis2.html> . The processed data used in
608 this work are available on reasonable request from the corresponding author.

609 **Author contribution:** NG processed and analyzed the data and wrote the original draft of
610 the manuscript. KA proposed the Idea, supervised this work and revised the manuscript.
611 YL helped in revising the manuscript.

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615 Physical Sciences Laboratory (PSL) for free accessibility to (NCEP/NCAR) reanalysis-II
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