

1 **Role of atmospheric aerosols in severe winter fog over Indo Gangetic Plains of India: a**
2 **case study**

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10

11 **Abstract**

12 Winter fog and severe aerosol loading in the boundary layer over north India, especially in the Indo-
13 Gangetic Plain (IGP), disrupts the daily lives of millions of people in the region. To understand better
14 the role of aerosol-radiation feedback on the occurrence, spatial extent, and persistence of winter fog;
15 and the associated aqueous chemistry in fog in the IGP, several model simulations have been
16 performed using the Weather Research and Forecasting model coupled with chemistry (WRF-Chem).
17 While WRF-Chem was able to represent the fog formation for the December 23-24, 2017 fog event
18 over the central IGP in comparison to station and satellite observations, the model underestimated
19 PM_{2.5} concentrations compared to the Central Pollution Control Board of India monitoring network.

20 While evaluating aerosol composition for fog events in IGP, we found that the WRF-Chem aerosol
21 composition was quite different from measurements obtained during the Winter Fog Experiment in
22 Delhi, with secondary aerosols, particularly chloride aerosol fraction being strongly underpredicted
23 (~66.6%). Missing emission sources (e.g., industry and residential burning of cow dung and trash) and
24 aerosol and chemistry processes need to be investigated to improve model-observation agreement. By
25 investigating a fog event on December 23-24, 2017 over central IGP, we found that the aerosol-
26 radiation feedback weakens turbulence, lowers the boundary layer height, and increases PM_{2.5}
27 concentrations and RH within the boundary layer. Factors affecting the feedback include loss of
28 aerosols through deposition of cloud droplets and internal mixing of absorbing and scattering aerosols.
29 Aqueous-phase chemistry increases the PM_{2.5} concentrations, which subsequently affects the aerosol-
30 radiation feedback by both increased mass concentrations and aerosol sizes. With aerosol-radiation
31 interaction and aqueous phase chemistry, fog formation began 1-2 hours earlier and caused a longer
32 fog duration than when these processes were not included in the WRF-Chem simulation. The increase
33 in RH in both the experiments is found to be important for fog formation as it promoted the growth of
34 aerosol size through water uptake, increasing the fog water content over IGP. The results from this

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35 study suggest that the aerosol-radiation feedback and secondary aerosol formation play an important
36 role in the air quality and the intensity and lifetime of fog over IGP, yet other feedbacks, such as
37 aerosol-cloud interactions, need to be quantified.

38

39 1 Introduction

40 The Indo-Gangetic Plain (IGP; 21°35'-32°28'N latitude. and 73°50'-89°49'E longitude) in the
41 northern part of the Indian subcontinent is one of the most densely populated and heavily polluted
42 regions in South Asia. The rapid population and economic growth in the IGP region over the last
43 decade have increased air pollution over this region. This is evident from the increasing trend in AOD
44 and NO_2 column concentration over India reported in recent studies (Dey and Di Girolamo, 2011;
45 Ghude et al., 2013; Krishna Moorthy et al., 2013), which has slowed and reversed only recently
46 (Sarkar et al., 2019). The high concentration of aerosols along the IGP and their adverse effects on
47 human health and the environment are increasing (Ghude et al., 2016). Consequently, more than 500
48 million people living in the IGP breathe air that exceeds the National Ambient Air Quality Standards
49 (NAAQS), which has reduced the life expectancy of the people (Debnath et al., 2022; Lelieveld et al.,
50 2015). Lelieveld et al., (2015) estimated a very high number of premature deaths (0.716 million per
51 year) linked to aerosols ($\text{PM}_{2.5}$), thus making Southeast Asia one of the largest regions affected by
52 premature mortality globally.

53 One of the major environmental concerns in the IGP is the urban air quality during winter,
54 especially over the mega-cities, e.g., Delhi, located in the north-western part of IGP (Ghude et al.,
55 2020; Jena et al., 2021; Sengupta et al., 2022). Several urban air pollution hotspots along the IGP
56 extend from northwest to east with monthly average $\text{PM}_{2.5}$ greater than $200 \mu\text{g m}^{-3}$ (~~NAAQS = $60 \mu\text{g m}^{-3}$~~ ,
57 24 hr average) in the winter season (Bharali et al., 2019; Krishna et al., 2019). IGP is dominated
58 mainly by fine mode particulates, especially over central to eastern IGP, during post-monsoon and
59 winter (Kumar et al., 2018). Biomass burning (agricultural waste burning, domestic heating, etc.) is an
60 important contributor to the observed high $\text{PM}_{2.5}$ loading over IGP during these seasons (Kulkarni et
61 al., 2020; Pant et al., 2015; Pawar and Sinha, 2022; Sharma et al., 2010; Yadav et al., 2020). Delhi is
62 affected substantially by the emissions from agricultural waste burning in the north-western states of
63 Punjab and Haryana during the post-monsoon (October-November) season (Badarinath et al., 2009;
64 Jethva et al., 2018; Kumar et al., 2021). Studies showed that $\text{PM}_{2.5}$ increased from $\sim 50 \mu\text{g m}^{-3}$ to as
65 high as $300 \mu\text{g m}^{-3}$ (Ojha et al., 2020), and AOD reached 0.98 with the presence of absorbing aerosols
66 (Singh et al., 2018) during the peak biomass burning in post-monsoon.

67 IGP experiences fog (both radiation and advection fog) every winter after the passage of the
68 synoptic wind system called the "Western Disturbances". The majority of fog events in the IGP during
69 December-January are radiation fog (Deshpande et al., 2023; Ghude et al., 2023), formed due to
70 radiative cooling of the surface. The number of low visibility days due to haze/fog formation has been

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72 increasing significantly (Ghude et al., 2017; Jenamani, 2007; Singh and Dey, 2012), impacting socio-
73 economic activities, e.g., aviation (Kulkarni et al., 2019). The increase in the intensity and regional
74 extent of fog over IGP is consistent with the increasing trend in aerosol concentration due to
75 increasing anthropogenic emissions (Sarkar et al., 2006; Syed et al., 2012).

76 Several factors control the formation and persistence of fog in the IGP, e.g., stable boundary
77 layer, low temperature, availability of moisture (supplied by the Western Disturbances and irrigation
78 activities), and the aerosol number and composition (Acharja et al., 2022; Dhangar et al., 2021). It has
79 also been suggested that the atmospheric rivers (moisture incursion from Arabian Sea) act as a source
80 of water vapor over IGP, which fuels the intensification of fog and haze (Verma et al., 2022) during
81 winter. The high aerosol concentration in the boundary layer influences fog formation (Gautam et al.,
82 2007; Safai et al., 2019) over the IGP by providing the needed cloud condensation nuclei (CCN) for
83 activation into fog droplets. In addition, aerosols induce surface cooling by reducing solar radiation at
84 the surface while warming the lower troposphere by absorption (Ding et al., 2016; Yu et al., 2002). A
85 reduction in surface-reaching solar radiation by ~19% has been reported during winter over Kanpur in
86 the IGP (Dey and Tripathi, 2007). The reduced solar flux affects the boundary layer stability and
87 depth by suppressing the thermals and thus further increasing the surface aerosol concentration via
88 aerosol-radiation feedback, which is very strong over the IGP (Bharali et al., 2019). Kumar et al.,
89 (2020) have shown that aerosol-radiation feedback significantly improves the accuracy of PM_{2.5} and
90 temperature forecasts in Delhi. Srivastava et al., (2018) reported that the direct aerosol forcing over
91 polluted regions is very large with values up to $-80.0 \pm 7.2 \text{ W m}^{-2}$ over the IGP in the winter season.

92 Aerosol-radiation interaction determines that the aerosol distribution is critical for the evolution
93 of fog (Bodaballa et al., 2022; Steeneveld et al., 2015), while microphysics is important for fog
94 formation and dispersal (Boutle et al., 2018; Maalick et al., 2016). Although the relationship between
95 the aerosol chemical composition and aerosol activation to CCN has not been fully understood yet,
96 studies have found that the chemical composition and mixing state of aerosols affect the
97 hygroscopicity (κ) of aerosols (Bodaballa et al., 2022; Ma et al., 2013; Moore et al., 2012; Zhang et
98 al., 2014a). Fog processes involve a complex interplay between local meteorology, radiation,
99 microphysics, and aerosol chemistry, making it difficult to understand the fog lifecycle (Acharja et al.,
100 2022; Maalick et al., 2016; Zhang et al., 2014b). There is considerable heterogeneity in the spatial and
101 temporal aerosol properties over IGP and the poor estimates of their mixing state. Therefore,
102 prediction of fog by weather models is still challenging with biases in fog's onset and dispersal
103 timings.

104 Previous studies have focussed on the impacts of meteorological conditions, topography, or
105 anthropogenic emissions on the poor air quality and intensification of fog during winter over IGP (e.g.
106 Hakkim et al., 2019). However, studies on the effect of feedback induced by the aerosols on the
107 meteorological conditions and thus on aerosol concentration are very limited over this region, except
108 for a few above-mentioned studies which discuss how the aerosol-radiation feedback favors haze and

109 fog during winter. Moreover, fog can provide a medium for aqueous-phase reactions. While several
110 earlier studies have reported an increase in secondary aerosols during fog over IGP, a sensitivity study
111 examining the impact of fog on aqueous phase chemistry has not yet been done over IGP.

112 In the present work, we aim to find the suitable chemistry/physics as well as the meteorology
113 initial/boundary conditions that lead to improved simulations of fog events in the Weather Research
114 and Forecasting model coupled with chemistry (WRF-Chem; (Fast et al., 2006; Grell et al., 2005;
115 Powers et al., 2017). We also explore the role of aerosol-radiation feedback on fog properties as the
116 high aerosol loadings in northern India can impact the heating rates, temperature inversions, and
117 boundary layer height. The role of aqueous chemistry on fog properties and vice-versa is also
118 investigated.

119

120 2 Methodology

121 Fog formed due to radiative cooling at the surface on both 23rd and 24th December 2017 over a
122 widespread region of the IGP (Fig. 1a, b). The fog region is located over an area with high PM_{2.5}
123 anthropogenic emissions (Fig. 1c). The IGP is a large region with varying meteorology and aerosol
124 characteristics, therefore, it is divided into three areas, northwest (NWIGP: latitude-longitude range,
125 27°N-32°N,75°E-79°N), central (CIGP: latitude-longitude range, 25°N-28°N,79°E-83°E), and east
126 (EIGP: latitude-longitude range, 24°N-27°N, 83°E-87°E) which are marked by the black rectangles in
127 Fig.1c. Although biomass burning and anthropogenic emissions dominate throughout the IGP during
128 post-monsoon and winter season, the north-westerly wind system results in the gradient distribution of
129 AOD over this region. The downwind regions, CIGP and EIGP are influenced by the long-range
130 transport from the NWIGP, resulting in high AOD with dominant fine particulates over CIGP and
131 EIGP, especially during post-monsoon and winter (Kedia et al., 2014; Kumar et al., 2018). Therefore,
132 representative stations from each region listed in section 2.2 are considered for the sensitivity
133 analyses.

134

135 2.1 Modeling

136 The WRF-Chem model version 4.0.3 is used for this study. Earlier studies have successfully
137 used WRF-Chem to predict fog (Pithani et al., 2019) and in the study of aerosol-radiation feedback on
138 air quality (Kumar et al., 2020; Bharali et al., 2019) and fog (Shao et al., 2023). The model domain is
139 centered at Delhi (77.1°E, 28.7°N) with 300 grid points in the east-west, 170 grid points in the south-
140 north direction (Fig. 1c), and 50 vertical eta levels with the model top at 50 hPa. The horizontal grid
141 spacing of the domain is 10 km, while the vertical grid spacing varies from higher resolution (~200 m)
142 in the boundary layer to coarser resolution (~1200 m) near the model top. We conduct three model
143 configurations (Table 1) for December 20-24, 2017 to identify the best configuration for
144 meteorological simulations. The three experiments have been designed with different combinations of
145 meteorological initial/lateral boundary conditions and planetary boundary layer (PBL) physics.

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146 Experiment 1 (EXP1) uses the National Centers for Environmental Predictions (NCEP) Final Analysis
147 (GFS-FNL; 1° x 1°, 6 hourly) meteorology data for initial and boundary conditions and the YSU
148 (Yonsei University; (Hong et al., 2006) PBL scheme. Experiments 2 and 3 (EXP2, EXP3) use ERA-
149 Interim Project (1.125° x 0.703°, 6 hourly) for meteorology initial and boundary conditions. EXP2
150 uses the YSU PBL scheme while EXP3 uses the ACM2 (Asymmetric Convective Model version 2)
151 PBL scheme. ACM2, is a hybrid of the original nonlocal closure (Pleim and Chang, 1992) and a local
152 closure eddy diffusion scheme (Pleim, 2007a, 2007b). The YSU PBL option was coupled with the
153 Noah LSM while ACM2 was coupled with Pleim-Xiu LSM. While YSU permits investigations of
154 both aerosol-radiation (AR) and aerosol-cloud interactions, aerosol-cloud interactions are not possible
155 when using the ACM2 PBL scheme because the ACM2 PBL scheme does not provide the exchange
156 coefficient for heat, which is required to calculate the activation fraction for mass and number for each
157 bin/mode. ACM2, on the other hand, has been shown to perform well for air quality in the IGP
158 (Mohan and Gupta, 2018), however, they tested the YSU and ACM2 schemes during the summer time
159 (1-15 June 2010) and focused on the evaluation of temperature, wind speed, PBL height, ozone, and
160 PM₁₀. To ensure that model captures all the relevant meteorological parameters including relative
161 humidity reasonably well during fog in winter, we designed EXP1, EXP2 and EXP3.

162 The advantage of Pleim-Xiu LSM (PX-LSM) is that it allows nudging of soil moisture and
163 temperature to improve the prediction of meteorology near the surface (Pleim and Gilliam, 2009;
164 Pleim and Xiu, 2003; Xiu and Pleim, 2001) which Noah LSM does not include. The PX-LSM
165 includes two-layer soil (0–1 and 1–100 cm) model, canopy moisture, and aerodynamic and stomatal
166 resistance. Ground surface (1 cm) temperature is calculated from the surface energy balance using a
167 force-restore algorithm for heat exchange within the soil. Although the two-layer approach in PX-
168 LSM is less detailed than the multilayer soil models such as the Noah LSM (four soil layers; Chen and
169 Dudhia 2001), it performs well with realistic initialization for soil moisture and through dynamic
170 adjustment in the model simulation where soil moisture is indirectly nudged according to differences
171 in 2-m temperature (T2) and 2-m relative humidity (RH) between the model and observation (Pleim
172 and Xiu, 2003). Soil moisture nudging adjusts the surface evaporation (direct soil surface evaporation,
173 vegetative evapotranspiration, and evaporation from wet canopies) which then affects the partitioning
174 of available surface energy into latent and sensible heat flux and thus reduces errors in T2 and 2-m
175 RH.

176 For EXP2, meteorological initial conditions were refreshed every 24 hours, while EXP3 was a
177 continuous run but soil moisture was nudged to the Era-Interim dataset to improve the prediction of
178 surface fluxes. All other physics and chemistry options are the same for all the experiments except the
179 surface physics option, which changes with the PBL scheme used. The deposition of cloud droplets is
180 an important moisture and aerosol sink during fog events. For all these simulations, the deposition
181 velocity of cloud droplets was reduced to 0.01 m s⁻¹ based on Stoke's Law and previous studies
182 (Katata et al., 2015; Tav et al., 2018) because its default value (0.1 m s⁻¹), is large.

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183 To examine the radiative effects of aerosols and aqueous phase chemistry additional
184 simulations have been done using the meteorological configuration in EXP3, with aerosol-radiation
185 (wFB) feedback plus aqueous chemistry (wAq.chem), without aerosol-radiation feedback (nFB) but
186 with aqueous chemistry, and without aqueous chemistry (noAq.chem) but with aerosol-radiation
187 feedback. The analysis has been done for the fog events on 23rd and 24th December 2017 as

188 Impact of radiation feedback=Parameters in wFB- Parameters in nFB

189 Impact of aqueous phase chemistry= Parameters in wAq.chem- Parameters in noAq.chem

190 Emissions used in the WRF-Chem simulations are from the EDGAR-HTAP v2 (*Emissions*
191 *Database for Global Atmospheric Research- Hemispheric Transport of Air Pollution; 0.1° x 0.1°*)
192 inventory for anthropogenic emissions and FINN v2.2 (*Fire INventory from NCAR; 1 km x 1 km*) fire
193 emission inventory (Wiedinmyer et al., 2011). Trash-burning emissions (Chaudhary et al., 2021) are
194 also included in the simulations. The model calculates the biogenic emissions online using MEGAN
195 v2.04 (*Model of Emissions of Gases and Aerosols from Nature*) (Guenther et al., 2006). The initial and
196 lateral boundary conditions for chemical constituents are from the global chemistry transport model
197 CAM-Chem (*Community Atmosphere Model with Chemistry*) (Emmons et al., 2020).

198 The MOZART (Model for Ozone and Related chemical Tracers) chemical mechanism
199 (Emmons et al., 2010) is used for gas-phase chemistry, which includes 85 gas-phase species, 39
200 photolysis, and 157 gas-phase reactions. It has been updated to include an explicit treatment of
201 aromatic compounds, HONO, C₂H₂, and isoprene oxidation scheme (Knote et al., 2014). The lumped
202 toluene used by Emmons et al., (2010) has been speciated into benzene, toluene, and lumped isomers
203 of xylenes (Knote et al., 2014). For this study, HCl emissions, transport, dry, and wet deposition are
204 represented. However, HCl gas-phase reaction is not included in MOZART.

205 The Model for Simulating Aerosol Interactions and Chemistry (MOSAIC) with four size bins
206 (0.039–0.156, 0.156–0.625, 0.625–2.500, and 2.5–10.0 μm dry diameters) coupled with MOZART
207 gas-phase chemistry is used (Fast et al., 2006; Zaveri et al., 2008). The bin sizes are defined by their
208 lower and upper dry particle diameters, so there is no transfer of particles between bins during water
209 uptake or loss. It is assumed that aerosols in each bin are internally mixed with the same chemical
210 composition while they are externally mixed in different bins.

211 The aerosol composition includes sulfate (SO₄²⁻), ammonium (NH₄⁺), nitrate (NO₃⁻), aerosol
212 water, sea salt (Na⁺, Cl⁻), methanesulfonate (CH₃SO₃), carbonate (CO₃²⁻), calcium (Ca⁺), black carbon
213 (BC), organic mass (OC), and unspecified inorganic species such as silica, inert minerals, and trace
214 metals lumped together as other inorganic mass (OIN). For OC, primary OC and secondary OC are
215 represented separately, where the latter is simulated using the volatility basis set (VBS) approach.
216 Reactive inorganic species such as potassium (K⁺) and magnesium (Mg⁺) are usually present in much
217 smaller amounts and are equivalent to Na⁺ since their sulfate, nitrate, and chloride salts are similar in
218 terms of their solubility in water.

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219 MOSAIC treats condensation and evaporation of trace gases to/from particles, nucleation
220 (new particle formation), and coagulation. Aerosol coagulation (Brownian) is based on (Jacobson et
221 al., 1994) and nucleation is based on (Wexler et al., 1994) parameterization of H₂SO₄-H₂O
222 homogeneous nucleation. Sulfate, nitrate, chloride, and ammonium aerosols are mainly formed
223 through oxidation and neutralization/condensation of gas precursors. Gas-phase sulfuric acid (H₂SO₄)
224 is produced by the gas-phase oxidation of SO₂ by OH and nitric acid (HNO₃) formation is via the
225 oxidation of NO₂ by OH. HCl is a primary emission product. The neutralization/condensation of
226 H₂SO₄, HCl, and HNO₃ with NH₃ produces ammonium such as ammonium sulfate (NH₄)₂SO₄,
227 ammonium bisulfate (NH₄HSO₄), ammonium chloride (NH₄Cl) and ammonium nitrate (NH₄NO₃),
228 respectively. The thermodynamic modules in MOSAIC for the dynamic gas-particle partitioning of
229 aerosols MTEM (Multicomponent Taylor Expansion Method) and MESA (Multicomponent
230 Equilibrium Solver for Aerosols) calculate the activity coefficient in aqueous phase aerosols and
231 compute the intraparticle solid-liquid phase equilibrium respectively (Zaveri et al., 2005, 2008). The
232 Adaptive Step Time-split Euler Method (ASTEM) coupled with MESA-MTEM dynamically
233 integrates the mass transfer equations.

234 Aqueous-phase chemistry uses a bulk water approach employing the Fahey and Pandis (2001)
235 mechanism. It calculates sulfate formation, formaldehyde oxidation, and non-reactive uptake of nitric
236 acid, hydrochloric acid, ammonia, and other trace gases (Chapman et al., 2009; Pye et al., 2020).
237 Aqueous-phase sulfate is produced via oxidation of SO₂ by H₂O₂, O₃, TMI (Transition metal Ion:
238 Fe(III), Mn(II)) catalyzed O₂ and NO₂. TMI concentrations are prescribed in the model to 0.01 μg m⁻³
239 for Fe(III) and 0.005 μg m⁻³ for Mn(II) (Martin and Good, 1991). The Fe(III) values are within the
240 range of water soluble iron in winter time aerosol reported in India (Kumar and Sarin, 2010). Wet
241 removal (scavenging), is represented by the (Neu and Prather, 2012) scheme for trace gases and Easter
242 et al., (2004) for aerosols.

243

244 **2.2 Observations**

245 To evaluate the model output, observations of aerosols and meteorology have been obtained
246 from several satellites as well as ground-based measurement platforms. To examine the aerosol
247 loading and spatial and temporal distribution, daily Level 2 Aerosol Optical Depth (AOD) retrievals
248 from the Moderate Resolution Imaging Spectroradiometer (MODIS) aboard Terra and Aqua satellites
249 are obtained at the spatial resolution of 10 km x 10 km (at nadir) pixel array. It provides aerosol
250 properties from the Dark Target (DT) algorithm applied over the ocean and dark land (e.g.,
251 vegetation) and Deep Blue (DB) algorithms over the entire land areas, including both dark and bright
252 surfaces. Each MOD04_L2 (Terra) / MYD04_L2 (Aqua) products are available at a 5-minute time
253 interval with an output grid of 135 pixels in width by 203 pixels in length.

254 The Indian National Satellites (INSAT-3D) in the geostationary orbit at inclinations of 82°
255 longitude provide an imager fog product (3DIMG_L2C_FOG) with a spatial resolution of 4 km every
256 30 min (www.mosdac.gov.in). For daytime, the visible channel observation is used to detect fog,
257 whereas thermal infrared is used to reduce false alarms such as medium/high clouds and snow areas.
258 INSAT 3D's 'day microphysics' data component analyzes solar reflectance at three wavelengths: 0.5
259 μm (visible), 1.6 μm (shortwave infrared), and 10.8 μm (thermal infrared). Night-time fog is derived
260 from TIR-1 (12.0 μm and 10.0 μm) and MIR (10.8 μm and 3.9 μm) channel brightness temperature
261 over the Indian region. INSAT-3D provides fog intensity varying from 1 to 4 indicating SHALLOW
262 for visibility > 600 m; MODERATE, DENSE, and VERY_DENSE, respectively for visibility varying
263 from 0 to 500 m (Banerjee and Padmakumari, 2020). If the visibility is greater than 700 m it indicates
264 no fog while visibility > 1000 m represents very clear skies. Validation of INSAT-3D fog products
265 over the IGP shows a 66%-68% probability of detection and a 10% false alarm rate. It also captures
266 the entire life cycle of fog from formation to dissipation. However, detecting fog during multilayer
267 clouds is still challenging with INSAT-3D (Arun et al., 2018; Chaurasia and Gohil, 2015; Chaurasia
268 and Jenamani, 2017).

269 Ground-based monitoring sites provide hourly data of relative humidity, surface temperature,
270 and wind speed measured by the Central Pollution Control Board, CPCB (<http://cpcb.nic.in>). Given
271 the data availability from CPCB stations, nine stations have been considered representing each region
272 of IGP, which include, Amritsar, IGI Airport (Indira Gandhi International Airport, Delhi), IHBAS
273 (Delhi), Dwarka (Delhi), RKP (Delhi) in the North-West IGP; Kanpur, Lucknow in Central IGP and
274 Patna, Muzaffarpur in East IGP.

275 In addition, measurements of several aerosols, trace gases, and meteorology at Delhi (IGI
276 Airport) from the Winter Fog Experiment (WiFEX) for the period December 10-31, 2017, have also
277 been used to validate the model output. The WiFEX, an initiative of the Ministry of Earth Sciences
278 (MoES), India, is a ground-based measurement campaign at the IGI Airport Delhi to understand fog's
279 physical and chemical features. Additional details of the WiFEX project and related publications can
280 be found in Ghude et al., (2017).

281

282 3 Meteorology Evaluation

283 Previous studies simulating fog highlight the importance of high model vertical resolution
284 (Pithani et al., 2019; Van Der Velde et al., 2010) for representing the fog formation and the growth of
285 the fog layer, model initialization (Yadav et al., 2022), initial relative humidity (Bergot and Guedalia,
286 1994; Pithani et al., 2020), and PBL schemes (Chen et al., 2020; Pithani et al., 2019). In the present
287 study, 2-m relative humidity (RH2), 2-m temperature (T2), and 10-m wind speed (WS) from WRF-
288 Chem have been evaluated using ground-based measurements from CPCB monitoring network and
289 WIFEX campaign for nine stations across the IGP. The comparison of WRF-Chem results with
290 observations shows that RH2 and T2 are sensitive to the choice of the meteorological initial and

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291 boundary conditions as illustrated by six stations in major cities (Fig. S1). WRF-Chem compares
292 better with the observations for simulations driven by the ERA-Interim reanalysis than with GFS-FNL
293 reanalysis since ERA-Interim provides more realistic RH2 than GFS-FNL (Figs. S2 a-f). For example,
294 RH2 from EXP1 (GFS) varies from 10 to 50%, while RH2 from EXP2 and EXP3 varies from 30 to
295 100%, which is closer to observation, especially for NWIGP and CIGP. For EIGP, RH2 from EXP1
296 (GFS) compares better than ERA-Interim, which overestimates the observed RH2. ERA-Interim and
297 YSU PBL scheme showed damping of RH2 continuously increasing the bias in RH2 with time (not
298 shown), which was corrected in EXP2 by refreshing meteorology every day at 00h UT during the
299 model simulation. In addition, maps of surface RH2 and T2 (Figs. S2 g-j) show that the GFS-FNL
300 dataset has lower relative humidity throughout the domain as compared to ERA-Interim. There are
301 differences in simulated 2-m temperature between these two datasets which are of smaller relative
302 magnitude compared to the RH2.

303 The GFS-FNL driven meteorology EXP1 has a warm bias in NWIGP and CIGP, especially during
304 night-time, while over EIGP, the model prediction agrees well with observations. EXP2 with the
305 ERA-Interim driven meteorology and YSU PBL scheme also shows good agreement between
306 modeled and observed T2 in EIGP. The ERA-Interim driven meteorology with the ACM2 PBL
307 scheme in EXP3 has a cold bias of up to 7°C over EIGP during daytime from 22nd to 24th December.
308 The wind speed evaluation shows that WRF-Chem is over-predicting wind speed. However, it is also
309 possible that some CPCB stations (e.g., Amritsar and RK Puram) have a wind speed low bias due to
310 the low measurement height and obstructions such as tall trees near the monitoring station as shown in
311 FigS3. WRF-chem in general overestimates wind speed and several earlier studies have reported this
312 bias in wind speed (e.g., Mohan and Gupta 2018; Pithani et al.,2019). Moreover, WRF-Chem does not
313 have the capability to represent building meteorology and parameterizes the effects of urban areas on
314 meteorology through roughness length, which likely leads to overestimation of wind speed. Note that
315 at other sites (e.g., over IGI-Delhi and Kanpur) the model measurement agreement is better.

316 The WRF-Chem performance has been statistically assessed against observation using the
317 Taylor Diagram (Taylor, 2001), which provides a statistical summary of how well the model output
318 agrees with the observation in terms of the Pearson correlation, their centered root-mean-square error
319 (RMSE) difference, and the ratios of their variances (Fig. 2).

320 The centered RMS difference, the correlation, and the standard deviation are related by the following
321 formula:

$$322 E'^2 = \sigma_o^2 + \sigma_m^2 - 2\sigma_o\sigma_m R$$

323 where R is the correlation coefficient between the model-simulated and observed fields, E' is the
324 centered RMS difference between the fields, and σ_m^2 and σ_o^2 are the variances of the model-simulated
325 and observed fields, respectively. The correlation(R), centered RMS difference (E') and standard
326 deviations of the model simulated and observed fields are calculated by the following formulas:

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327
$$R = \frac{\frac{1}{N} \sum (M_n - \bar{M})(O_n - \bar{O})}{\sigma_m \sigma_o} \quad (1)$$

328
$$E'^2 = \frac{1}{N} \sum [(M_n - \bar{M}) - (O_n - \bar{O})]^2 \quad (2)$$

329
$$\sigma_m^2 = \frac{1}{N} \sum (M_n - \bar{M})^2 \quad (3)$$

330
$$\sigma_o^2 = \frac{1}{N} \sum (O_n - \bar{O})^2 \quad (4)$$

331 where the overall mean of a field is indicated by an overbar.

332 The percentage bias has also been included to further evaluate the WRF-Chem results. In Fig.
 333 2, better agreement of WRF-Chem results with observations are shown by the marker's proximity to
 334 the "OBS" dashed black line. The WRF-Chem RH has a good correlation for all three experiments
 335 with $r > 0.75$ at all the locations in IGP for all the experiments. However, the RMSE (shown by red
 336 dashed contours) and the standard deviations are larger for the EXP1. The relative bias is also large
 337 ($>20\%$) for EXP1 (GFS-FNL) compared to EXP2 and EXP3 which lie closer to the dashed black line
 338 indicating that the simulated RH variations are similar to observations. For all the experiments, WRF-
 339 Chem T2 agrees well with observations with a correlation between 0.8 and 0.95. The points are
 340 concentrated near the dashed line showing a low RMSE and standard deviation for T2, signifying a
 341 good agreement of simulated T2 with observation in terms of temporal variation but the T2 relative
 342 bias is large for EXP1 ($>20\%$). The RMSE and relative bias for EXP1 are larger for several of the
 343 stations. The temporal variability of T2 and RHs predicted well for all the combinations of inputs
 344 (Fig. S1), however, the accuracy of simulated T2 and RH is sensitive to the choice of meteorological
 345 initial/boundary conditions. WRF-Chem predicted RH and T2 agree better with observations when
 346 initialized with ERA-Interim meteorology than with GFS-FNL.

347 The WRF-Chem runs driven by ERA-Interim with YSU (EXP2) and ACM2 PBL (EXP3)
 348 schemes predicted the surface meteorology better over the IGP than the WRF-Chem run driven by
 349 GFS (EXP1). By examining the modeled cloud water content in the lowest model level with the
 350 INSAT-3D satellite fog intensity for the 23rd and 24th December 2017 (Fig. 3), it is apparent that
 351 WRF-Chem with the ACM2 PBL scheme compared qualitatively well with observations obtained
 352 from INSAT-3D satellite in terms of fog coverage over CIGP, while the WRF-Chem run with the
 353 YSU PBL scheme did not produce widespread fog. However, there is also fog over EIGP in WRF-
 354 Chem with the ACM2 PBL scheme although it is not observed by the satellite. This is because the
 355 model has a cold bias in T2 and a high surface RH over East IGP with ACM2 PBL and Pleim-Xiu
 356 surface scheme as discussed earlier, which favors the formation of fog in this region. The time series
 357 in Fig. 4 shows that EXP3 is capable of predicting the duration of fog on 23rd and 24th December.
 358 There is a data gap from INSAT 3D observations because it is unable to capture fog during daytime in
 359 the presence of mid and high-level clouds.

360 In conclusion, EXP3 is the best configuration for predicting fog formation where the ERA-
 361 Interim meteorology, the ACM2 PBL and surface schemes, and soil moisture nudging is used in the

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363 WRF-Chem simulation. Therefore, the evaluation of predicting AOD, surface aerosol concentrations,
364 and aerosol composition as well as analysis of the impact of aerosols on fog formation uses the EXP3
365 configuration.

366

367 **4 Aerosol Evaluation**

368 Aerosol is an important factor in correct prediction of fog (Maalick et al., 2016; Stolaki et al.,
369 2015) as the number of fog droplets depends on the aerosol size distribution and concentration.
370 Aerosols as CCN can affect the liquid water content in fog and therefore an increase in aerosol
371 concentration can significantly affect fog lifetime (Stolaki et al., 2015; Zhang et al., 2014b). AOD
372 retrievals from the MODIS satellite have been used to validate the modeled AOD (Fig. 5). It is
373 observed that the model captures several important features of the MODIS retrieved AOD spatial
374 distribution but at the same time somewhat struggles to reproduce the observed AOD magnitude in
375 some parts of the domain. One possible reason for the underestimation would be the EDGAR-HTAP
376 emission inventory, which has a low bias for residential sector PM_{2.5} emissions in India (Sharma et al.,
377 2022). For instance, the model successfully predicts high aerosol loading seen by MODIS on 20 and
378 21 December over CIGP and EIGP. This is the region with dense fog both in model and observation.
379 Higher AOD (>0.5) over CIGP and EIGP can be attributed to the accumulation of aerosols that are
380 transported by north-westerly winds to these regions from NWIGP (Dey and Di Girolamo, 2011; Jain
381 et al., 2020; Jethva et al., 2018; Kumar et al., 2018; Yadav et al., 2020). However, WRF-Chem
382 underestimates AOD over the NWIGP (AOD<0.3) throughout the simulation period and during 23-24
383 December over CIGP and EIGP where the latter may be related to enhanced scavenging of aerosols by
384 fog droplets.

385 The west to east gradient in aerosol loading over IGP is consistent with surface PM_{2.5}
386 distribution (Fig. 6a). Surface PM_{2.5} concentration is highest in EIGP (>100 µg/m³) and it decreases
387 gradually towards NWIGP (~60-80 µg/m³). The time series of PM_{2.5} from CPCB measurements and
388 the model at stations representative of each region in IGP shows that simulated PM_{2.5} compares well
389 with observation in terms of day-to-day variation over most of the locations in the IGP (Fig. 6 b-e).
390 The comparison is good over Amritsar (an NWIGP location), where PM_{2.5} is mostly primary aerosols
391 from local emissions e.g., residential heating related biomass burning. Agricultural waste burning is at
392 its peak during post monsoon months (Oct-Nov), whereas during winter burning for residential
393 heating increases and the stable boundary layer confines these emissions near the surface (Kumar et
394 al., 2021; Pawar and Sinha, 2022). PM_{2.5} at Amritsar shows a bimodal distribution with morning and
395 evening peaks whereas it is absent in the model likely due to the absence of diurnal variations in the
396 WRF-Chem anthropogenic emissions.

397 A statistical analysis (Table S1) shows a minimum mean bias for PM_{2.5} at Amritsar (-2.2%) while in
398 other stations it ranges from 48 to 53% similar to the reported range of model bias (underestimated by
399 40–60%) in winter over IGP by earlier studies (Bran and Srivastava, 2017; Ojha et al., 2020). RMSE

400 values range from 41 to 138 $\mu\text{g}/\text{m}^3$ (normalized RMSE~0.4 to 0.7) comparable to the reported values
401 by these studies. The Pearson correlation coefficient (r) for the simulated and observed day-to-day
402 variation in $\text{PM}_{2.5}$ lies between 0.4 and 0.7 for all the stations in Fig. 6 except at Patna which lies
403 within the range in these studies. Poor correlation at Patna is due to the loss of $\text{PM}_{2.5}$ during fog in the
404 model as discussed earlier.

405 At Delhi, the daily variations are predicted well although WRF-Chem underestimates $\text{PM}_{2.5}$
406 observations during the first 4 days. Delhi experiences severe air pollution and haze with high PM
407 loading ($> 500 \mu\text{g m}^{-3}$) (Bharali et al., 2019). The model is successful in predicting the high $\text{PM}_{2.5}$
408 episode on the 24th of December, but WRF-Chem underpredicts the SO_4^{2-} , NH_4^+ , NO_3^- and Cl
409 concentrations (Fig. 7). Although simulated SO_2 and NH_3 are comparable with observation, sulfate,
410 and ammonium are underestimated in the model. SO_4^{2-} is underestimated by $\sim 9 \mu\text{g m}^{-3}$, while NH_4^+ ,
411 NO_3^- and Cl are underestimated by $\sim 30 \mu\text{g m}^{-3}$, $\sim 19 \mu\text{g}/\text{m}^3$ and $\sim 40 \mu\text{g}/\text{m}^3$ on average, respectively. In
412 addition, the WRF-Chem model results show that a large percentage of $\text{PM}_{2.5}$ is classified as “other
413 inorganics”, which is usually dominated by $\text{PM}_{2.5}$ other than BC and OC. This leads to the
414 underestimation of $\text{PM}_{2.5}$ over Delhi. Studies report very high chloride over the IGP with values
415 exceeding $100 \mu\text{g m}^{-3}$ (Lalchandani et al., 2021) during winter emitted from increased trash burning
416 and industrial emissions (Pant et al., 2015; Patil et al., 2013). WRFChem incorporates trash-burning
417 emissions which include HCl emissions from Chaudhury et al.,(2021) for this study however, the
418 inventory contains annual emissions and fails to resolve the seasonality of trash-burning emissions as
419 identified by Nagpure et al., (2015). They suggested almost all the waste-burning emissions in
420 neighbourhoods with higher socioeconomic status in Delhi occur due to the use of waste as cheap
421 heating fuel by individuals such as night watchmen and pavement dwellers. Chaudhary et al., (2021)
422 considers waste burning that occurs due to lack of collection infrastructure, and at landfills and,
423 therefore, shows a concentration of waste burning emissions around the periphery of Delhi but low
424 waste burning emissions in the relatively prosperous city centre. In addition, emissions from other
425 sources (e.g., industries) are unaccounted for in the model which likely leads to the underestimation in
426 modeled chloride.

427 Over the CIGP and EIGP, the underestimation in $\text{PM}_{2.5}$ is mostly observed at the east IGP
428 locations during the dense fog. It is well known that the hygroscopic aerosols grow in size and are
429 deposited to the surface during fog (Gupta and Mandariya, 2013; Kaul et al., 2011). $\text{PM}_{2.5}$ shows an
430 increase initially with the onset of fog and then it decreases as the aerosols grow and get deposited
431 through fog droplets. A two order higher deposition rate (Fig. 6 f, g) during fog compared to the
432 deposition rate of dry aerosol results in the lower $\text{PM}_{2.5}$ over CIGP and EICP during fog events.

433 Previous studies have reported that models tend to underestimate the AOD observation (David
434 et al., 2018; Pan et al., 2015) during the post-monsoon and winter when agricultural waste burning and
435 anthropogenic emissions dominate. While anthropogenic emissions include a contribution from the
436 residential sector, the emissions from small-scale burning for residential heating over IGP especially

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437 during winter are likely underestimated in the current emission inventory (Sharma et al., 2022). This
438 leads to an underestimation of aerosol concentration in the model. Other possible causes for the
439 underestimation are the biases in the simulated meteorology (Govardhan et al., 2015; Kumar et al.,
440 2015; Pan et al., 2015) which affects the aerosol concentration. We corrected some of the biases in
441 meteorology as discussed earlier however there are still residual biases in the simulated meteorology
442 e.g., overestimation of wind speed by WRF-Chem. We also observe underestimation of secondary
443 aerosols over NWIGP which contribute significantly to the aerosol loading over IGP. Secondary
444 aerosol formation is substantial over CIGP and EIGP in the model compared to NWIGP which will be
445 discussed in a later section. The underestimation of $PM_{2.5}$ could also be linked to the uncertainty in the
446 model's chemistry scheme to simulate the secondary aerosols due to missing chemical processes or
447 due to underestimation of sulfur oxidation at different RH levels (Acharja et al., 2022; Pawar et al.,
448 2023; Ruan et al., 2022). Moreover, several modeling studies have shown significant improvements
449 in forecasting surface $PM_{2.5}$ by assimilation of satellite AOD and $PM_{2.5}$ (Ghude et al., 2020; Jena et al.,
450 2020; Kumar et al., 2020) suggesting the importance of correct initialization of the model in
451 simulating aerosols over IGP.

452

453 **5 Effect of Aerosol Radiation feedback**

454 Interactions of aerosols with radiation affects temperature and surface heat fluxes, thereby
455 weakening the turbulence in the PBL and stabilizing the boundary layer height (Fig. 8b) compared to
456 the clean environment (Fig. 8a). In the presence of well mixed aerosols within the PBL, the radiative
457 effect of aerosols lowers the noontime PBL height (Fig. 8b). However, the presence of absorbing
458 aerosols in the PBL warms the air and changes the thermodynamics. Three cases are shown in Fig.
459 8(c-e) where increases of scattering aerosol concentrations at the top of PBL (Fig. 8c) increases
460 scattering of radiation by the aerosol layer and reduces the surface reaching solar radiation similar to
461 Fig. 8b. Higher concentrations of absorbing aerosols at the top of PBL (Fig. 8d) warms the air above
462 the boundary layer and strengthens the capping inversion stabilizing the PBL and suppressing its
463 growth. The shallow PBL and weakened daytime vertical mixing confines aerosols and water vapor
464 near the surface and worsens the air quality of a region. The aerosols trapped in the stagnant PBL
465 further affects the radiation flux at the surface and creates a positive feedback loop wherein the PBL is
466 continually suppressed until interrupted by some synoptic weather phenomenon, such as the western
467 disturbances in the IGP. On the other hand, higher concentration of absorbing aerosols within the PBL
468 (Fig. 8e) warms the air in the PBL and this results in the higher PBL height. The raised PBL decreases
469 the aerosol concentration near the surface which is termed as a negative feedback effect.

470 The aerosol radiation feedback can affect shortwave heating rates (SWHR). The high aerosol
471 loading over the IGP (Fig. 6 and Fig. 7) allows the AR feedback to reduce the PBL height by more
472 than 140 m throughout the IGP compared to the surrounding region with AR feedback (Fig. 9a). The
473 difference in PBL height with and without aerosol radiation feedback is largest during noontime. The

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474 suppressed PBL is due to the decrease in the surface heating flux and the consequent weakening of
475 turbulence in the PBL. The surface solar radiation flux (SWF) decreases by 5-35 % while the surface
476 latent heat (LH) and sensible heat (HFX) fluxes decrease by 5-35 % and 10-60 %, respectively (Fig.
477 S3). The stable, shallow PBL reduces the vertical mixing of aerosols and moisture and confines them
478 near the surface, resulting in increased $PM_{2.5}$ concentrations and RH near the surface with AR
479 feedback (Fig. 9). Although T2 should decrease with the reduction in surface SWF, T2 shows mixed
480 signals with both cooling and warming over IGP. While surface cooling is observed over NWIGP and
481 EIGP, T2 increases with AR feedback over most of CIGP. The response of AR feedback to T2 varies
482 in these three regions probably due to differences in the distribution and types of aerosols and the
483 presence of fog. Increase in surface concentration of $PM_{2.5}$ occurs more over NWIGP and EIGP with
484 increase in BC and OIN over NWIGP, and sulfate aerosol over EIGP which results in the surface
485 cooling due to positive AR feedback in these two regions. Over the CIGP, the AR feedback causes a
486 depletion of surface $PM_{2.5}$ (Fig. 9d), which is likely due to their hygroscopic growth, and then dry
487 deposition (average dry deposition flux of $PM_{2.5}$ =331 $\mu\text{g}/\text{m}^2/\text{hr}$ with AR feedback and 282 $\mu\text{g}/\text{m}^2/\text{hr}$
488 without AR feedback) in dense fog. The increase in RH with AR feedback favours the growth of
489 aerosols in size by the uptake of water

490 Examining further, the time variation of the changes in PBL height, T2, and RH between the
491 simulations with and without aerosol-radiation feedback (Fig. 9g) shows an increase in T2 while the
492 surface fluxes, sensible heat flux, latent heat flux, and downward shortwave radiation flux decrease
493 over CIGP (Fig. 9h). AR feedback affects mostly the lower atmosphere at multiple levels; however,
494 our finding suggests that the decreased shortwave radiation flux decreases the surface fluxes and thus
495 the turbulence in the boundary layer resulting in a reduced PBL height on both days. Figure 9 g and h
496 clearly show a decrease in HFX and LH following the decrease in SWF. Moreover, we observe that
497 the PBL height is sensitive to latent heat flux likely due to its strong dependence on moisture
498 availability (Xiu and Pleim, 2001; Zhang and Anthes, 1982).

499 The impact of AR feedback on T2 depends on factors such as the presence of absorbing
500 aerosols and their vertical distribution via heating or increased SWF (as observed in CIGP, Fig. S4).
501 Absorbing aerosols in WRF-Chem include BC and OIN (other inorganic aerosols), which both
502 increase near the surface (Fig. 9e, Fig. S5) due to their confinement in the stable PBL. Some areas in
503 the fog-affected region show a decrease in BC as well as SO_4^{2-} likely due to increased dry deposition
504 in fog water as discussed earlier in this section for $PM_{2.5}$. As a result, AR feedback changes the
505 absorbing to scattering ratio of aerosols over IGP indicated by the decrease in SSA (Single Scattering
506 Albedo; Fig. S6). In EIGP, sulfate concentrations is larger with AR feedback than without AR
507 feedback with time periods where the difference is $>1 \mu\text{g}/\text{m}^3$ (Fig.11). The BC concentration changes
508 are small ($<0.5 \mu\text{g}/\text{m}^3$) in the EIGP, resulting in a higher SSA near the surface with AR feedback in
509 EIGP. In the CIGP, BC concentrations increase while sulfate aerosols decrease within the PBL with
510 AR feedback (Fig.11) compared to the simulation without AR feedback. A decrease in SSA is seen for

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511 the CIGP throughout the boundary layer while in EIGP the decrease occurs near the top of the PBL;
512 difference in SSA due to AR feedback is negligible in NWIGP. Also contributing to the higher SSA in
513 EIGP is the increase in RH (Fig. 9) due to AR feedback favoring the growth of aerosols in size by
514 uptake of water and the production of secondary aerosols such as SO_4^{2-} and NH_4^+ .

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515 A similar observation has been made by Ramachandran et al., (2020) where SSA decreases
516 with increasing altitude due to absorbing carbonaceous aerosols at higher elevations which contributes
517 $\geq 75\%$ to the aerosol absorption over IGP. Increased shortwave heating (Fig. 10) is probably caused by
518 the increased absorbing aerosols near the surface which overwhelms the surface cooling due to
519 reduced shortwave radiation at the surface.

520 The increase in 2-m RH is substantial over CIGP on 24th December (Fig. 9g) compared to the
521 previous day following the decrease in PBL height which constrains the moisture near the surface.
522 The decrease in RH by 2% or more when aerosol-radiation feedback is included compared to no
523 aerosol-radiation feedback is likely due to increase in T2. However, the increase in RH in the
524 afternoon associated with a decrease in LH and PBL height is important for the air to saturate which
525 then favors the formation of fog in a polluted environment. Note that the increase in T2 with AR
526 feedback is very small ($< 0.5^\circ\text{C}$) which reduces further after noon ($\sim 12:30$ pm IST) on both days.

527 Another important factor that can affect the extent of change in PBL height is the distribution
528 of aerosols in the vertical (illustrated in Fig. 8). The pressure-time cross-sections of differences in T,
529 $\text{PM}_{2.5}$, BC, and SO_4^{2-} between aerosol radiation (AR) feedback (wFB) and no aerosol radiation
530 feedback (nFB) for three regions, NWIGP, CIGP, and EIGP are shown in Fig. 11. The difference in
531 the PBL height reaches a maximum with the AR feedback during midday (12:30-15:30 IST). Increase
532 in temperature in the boundary layer is observed with AR feedback particularly at the upper PBL in all
533 the regions of IGP. This induces a temperature inversion resulting in a stable and suppressed PBL. In
534 all the regions the decrease in PBL height (100-200 m) is larger on 24th December compared to 23rd
535 December. The difference in the PBL height on 23rd and 24th December with AR feedback on these
536 days is possibly controlled by the aerosol distribution during the previous day or early morning on the
537 same day. For example, in all the regions an increase in $\text{PM}_{2.5}$ is observed the previous night (23:30
538 onwards) till $\sim 11:30$ of December 24, with increased BC over NWIGP and CIGP whereas both BC
539 and SO_4^{2-} over EIGP. The increased $\text{PM}_{2.5}$ concentrations suppress the development of the PBL after
540 sunrise with AR feedback on December 24 compared to that on December 23, leading to the observed
541 differences in ΔPBL height on these two days. Increase in BC concentrations in NWIGP and CIGP are
542 found above the PBL on 24th December whereas BC concentrations decrease within the PBL. This BC
543 concentration gradient creates a temperature inversion, for example between 10:30-14:30 IST. The
544 increase in BC warms the air in the PBL; however, the warming is not strong enough to cause
545 negative feedback over CIGP. On 23rd December a small increase in BC is uniform throughout the
546 PBL, while there is a decrease in SO_4^{2-} concentrations, resulting in a warmer PBL (Fig. 11) with AR
547 feedback.

548 In EIGP, BC distribution is similar to that in CIGP with AR feedback while there is a
549 substantial increase in sulfate aerosol in the PBL. This results in the strongest extinction in EIGP as
550 evident from the largest difference in PBL height and surface cooling with AR feedback among the
551 three regions. Although Δ PBL is small on 23rd December, it still results in the accumulation of
552 aerosols during night-time (~23:30 pm onwards) which further strengthens the AR feedback effect the
553 next day in NWIGP and CIGP. Thus, AR feedback stabilizes the PBL, increases PM_{2.5} and RH in the
554 PBL making conditions favourable for persistence of fog over IGP.

555

556 6 Effect of Aqueous phase chemistry

557 In this section we discuss the impact of aqueous phase chemistry on aerosol composition and
558 its interaction with meteorology. There is a considerable difference in the surface concentration of
559 PM_{2.5} ($>16 \mu\text{g m}^{-3}$) in the absence of aqueous chemistry over CIGP and EIGP where fog occurs (Fig.
560 12a) while the difference is negligible over NWIGP where fog does not occur. This is due to the
561 formation of secondary aerosols through aqueous phase chemistry and the hygroscopic growth of
562 aerosols during fog in these regions with the inclusion of aqueous chemistry in the model. In the
563 region between CIGP and EIGP (83E-84E; marked by the box in Fig. 12a), PM_{2.5} concentration is less
564 in the simulation with aqueous-phase chemistry than without aqueous-phase chemistry because
565 deposition of fog water aerosols to the surface increases as the fog thickens (Fig. 13, Fig. S7). Figure
566 13 shows the relation between formation of secondary aerosols, deposition flux of PM_{2.5}, and fog with
567 and without aqueous phase chemistry. During the fog event, the secondary aerosols (SO_4^{2-} , NH_4^+)
568 increase significantly by 4-10 $\mu\text{g m}^{-3}$ due to aqueous phase chemistry adding to the PM_{2.5} burden over
569 IGP. The intensity of fog is high around midnight December 24-25 compared to that on 23rd and 24th
570 (1:30-11:30 IST)) which increases the dry deposition flux of PM_{2.5} causing a sharp drop in the PM_{2.5}
571 concentration on 24th December compared to the previous night's fog event. The observed change in
572 PM_{2.5} over a region is the net result of the formation of secondary aerosols and its deposition with fog
573 droplets.

574 The composition distribution of PM_{2.5} (Fig. 12b) has a similar distribution for the simulations
575 with and without aqueous phase chemistry over NWIGP where fog did not occur. The primary
576 aerosols are higher (BC $> 9\%$, OC ~ 16 -30%, OIN $> 50\%$), than the secondary aerosols ($<5\%$). While
577 the model requires fog for accelerated formation of secondary inorganic aerosol, experimental data
578 (Fig. 7) supports significant formation of secondary inorganic aerosol at elevated RH levels even in
579 haze aerosol (Acharja et al., 2022). On the other hand, the central and east IGP stations are fog-
580 covered and therefore, there is an increase in secondary aerosols especially SO_4^{2-} and NH_4^+ when
581 aqueous phase chemistry is included in the simulation. SO_4^{2-} is chemically produced via aqueous
582 phase chemistry in cloud water, hence the abrupt increase whereas NH_4^+ maintains a gas-aerosol and
583 gas-cloud equilibrium with NH_3 and SO_4^{2-} via neutralizing the drop or aerosol. NO_3^- is high in the
584 model compared to SO_4^{2-} and NH_4^+ and it decreases by ~ 1 -2 % with aqueous phase chemistry. We

585 observe a small increase in NO_3^- during fog, however it drops as fog intensifies, more rapidly than that
586 without aqueous phase chemistry likely due to increase in dry deposition. This results in lower
587 average NO_3^- to $\text{PM}_{2.5}$ ratio with aqueous phase chemistry. Moreover, NO_3^- is high over the fog
588 covered CIGP and EIGP compared to NWIGP suggesting that transport and chemistry of NO_x in
589 CIGP and EIGP produce more HNO_3 . Aerosol NO_3^- is also in equilibrium with HNO_3 and it is formed
590 only if excess NH_3 is available beyond the sulfate neutralization. Thus, NH_4^+ and NO_3^- changes are
591 likely due to changing the partitioning between gas and liquid based on the production of sulfate.

592 $\text{PM}_{2.5}$ is mostly composed of organic aerosols (OA) over CIGP and EIGP (Lalchandani et al.,
593 2021; Srinivas and Sarin, 2014) whereas $\text{PM}_{2.5}$ is OIN (dust) and OA over NWIGP (Ram et al.,
594 2012a; Sharma and Mandal 2023). Although observational studies report Cl^- as one of the largest
595 contributors (12-17%) to $\text{PM}_{2.5}$ after the organics (Lalchandani et al., 2021; Pant et al., 2015) during
596 winter, Cl^- is largely underestimated by the model as discussed in section 4 and contributes only ~3%.
597 A small increase (2-4%) in secondary organic aerosols (SOA) from glyoxal production in aerosols
598 occurs for the simulation with aqueous phase chemistry included during intense fog, suggesting there
599 are feedbacks between cloud chemistry (without glyoxal aqueous chemistry) and aerosol chemistry.
600 However, similar to NO_3^- , average SOA (ASOA (anthropogenic)+BSOA (biogenic) + GlySOA)
601 shows a decrease when aqueous phase chemistry is included. SOA contributes significantly to organic
602 aerosol loading over IGP (Kaul et al., 2011; Mandariya et al., 2019).

603 The WRF-Chem results on aerosol composition during fog behave similarly to observational
604 studies. For example, Ram et al., (2012a) reported an increase of EC, OC, and WSOC concentrations
605 by ~30% during fog and haze events at Allahabad, a location in the Central IGP, and a marginal
606 increase of these constituents at Hisar (NWIGP). Several studies report an increase in inorganic ions
607 (NH_4^+ , NO_3^- , and SO_4^{2-}) during fog over IGP and elsewhere (Gundel et al., 1994; Ram et al., 2012a).
608 Recent studies suggest that a significant fraction of atmospheric particulate matter in the IGP is
609 comprised of carbonaceous aerosol (~30–35% of the PM) and water-soluble inorganic species (~10–
610 20% of the PM) during October–January when emissions from biomass burning (including residential
611 heating) are dominant over IGP (Ram et al., 2012b; Rengarajan et al., 2007; Tare et al., 2006).

612 Both the simulations with and without aqueous-phase chemistry include the AR feedback. The
613 aqueous chemistry increases the mass of $\text{PM}_{2.5}$ and the size of the aerosols, both of which contribute to
614 AR feedback, thus increasing RH and PBL stability. The increase in RH also saturates the air,
615 promotes aerosol growth by water uptake, and thus favors fog formation. Since the secondary
616 inorganic aerosols are scattering aerosols, the increased scattering of radiation further reduces the
617 solar radiation reaching the surface (Fig. 14a). Over CIGP the presence of higher aerosol loading
618 reduces the T2 during daytime, particularly on the 24th of December which then reduces the PBL
619 height and increases RH near the surface (Fig. 14b). These conditions favor fog formation over the
620 CIGP. Further, the fog water content with aqueous-phase chemistry is higher than that without
621 aqueous-phase chemistry on 24th December post-midnight (Fig. 13b). This is likely due to saturation

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622 of air due to increase in RH and lower T2, induced by the AR feedback caused by the increase in
623 PM_{2.5}. Although the difference in T2 is small (<0.4), favourable conditions mentioned above are
624 conducive to fog formation. [Because aqueous chemistry increases sulfate concentrations, the size of](#)
625 [the aerosols also increase. The increased aerosol size, which can grow further by water uptake, also](#)
626 [impacts the solar radiation reaching the surface, affecting fog formation and dissipation.](#)

627

628 **7 Effect of AR feedback and aqueous chemistry on the duration of fog**

629 Aerosol and its radiative effects impact fog characteristics, including the fog liquid water content
630 (LWC), the fog lifetime over a region and hence its spatial and temporal distribution. Variations of fog
631 LWC in WRF-Chem contrast the fog in the CIGP and EIGP (Figure 15) as well as among the three
632 experiments (with aqueous chemistry plus AR feedback, with aqueous chemistry without AR
633 feedback, and without aqueous chemistry but with AR feedback). [WRF-Chem does not simulate fog](#)
634 [over NWIGP in the model for the study period. In Figure 15, only foggy grid points are considered for](#)
635 [the first fog event on 23-24 December. The LWC is 5-15% higher with AR feedback than without AR](#)
636 [feedback and without aqueous phase chemistry for both CIGP and EIGP. The interquartile range is](#)
637 [larger for the simulation with and without AR feedback than without aqueous phase chemistry in](#)
638 [CIGP showing large variability in the LWC. On the other hand, in EIGP the variability in LWC is](#)
639 [greater in the simulation with AR feedback compared to the other two experiments.](#)

640 [The formation and dissipation times of the two fog events for the three experiments are listed](#)
641 [in Tables 2 and 3 for CIGP and EIGP. The 23-24 December fog starts forming two hours earlier and](#)
642 [the 24-25 December fog forms one hour earlier in both CIGP and EIGP with AR feedback than](#)
643 [without AR feedback. In the simulation without aqueous phase chemistry, fog formation is delayed by](#)
644 [an hour or two compared to the simulation with aqueous chemistry plus AR feedback in CIGP. In](#)
645 [EIGP the 23-24 December fog forms at the same time with AR feedback and without aqueous phase](#)
646 [chemistry while the 24-25 December fog is delayed by an hour without aqueous phase chemistry. Fog](#)
647 [dissipation usually occurs after sunrise when the shortwave radiative warming at the surface warms](#)
648 [the air, which results in PBL mixing. In addition, absorbing aerosols like BC affect fog dissipation by](#)
649 [increasing the radiative heating in and above the fog. We find an increase in BC and shortwave](#)
650 [heating in the PBL with AR feedback \(Fig. 10,11\) and warming over CIGP with AR feedback. Fog](#)
651 [intensity starts to decrease after 01:00 UTC \(06:30 IST\), however, in our study, we find that the fog](#)
652 [dissipates completely in the afternoon \(~10:00 UTC or 15:30 IST\) for both the simulations with AR](#)
653 [feedback and no aqueous chemistry while an hour later without AR feedback in CIGP. Fog dissipation](#)
654 [is delayed in EIGP with AR feedback compared to that without AR feedback and without aqueous](#)
655 [phase chemistry. In both the regions, fog lifetime increases with AR feedback. All the stations,](#)
656 [however do not show the same pattern, for example, the 23-24 December fog in Lucknow forms and](#)
657 [dissipates at the same time for simulations with AR feedback and without aqueous phase chemistry,](#)
658 [and the 24-25 December fog forms later with AR feedback than without AR feedback. Patna shows no](#)

Commented [CB25]: Time format corrected as suggested by Reviewer1, Minor comment 12

659 difference in the 24-25 December fog formation in all the three experiments. To gain better insights on
660 the fog timing, we recommend that simulations at higher spatial and temporal resolutions be
661 performed to represent better the fog dynamics at point locations. Furthermore, there are other
662 important factors to consider, e.g., improved emissions, better simulations of aerosol chemical
663 composition, and evaluation of aerosol deposition.

664 The AR feedback and aqueous-phase chemistry have the potential to impact aerosol-fog
665 interactions. We can learn about the effect of the aerosol-radiation interactions on CCN concentrations
666 because the WRF-Chem model calculates the CCN concentrations at different supersaturations as a
667 diagnostic output. We compare CCN at 0.02% supersaturations, a value typical of fog, among the
668 three experiments. For the 23-24 December fog in CIGP, hourly CCN concentrations are ~10% higher
669 for the simulations with AR feedback with or without aqueous chemistry than with no AR feedback
670 (Figure S8) during the first 8 hours of the fog event (16:00-24:00 IST 23 December). Surprisingly, the
671 simulation with no aqueous chemistry has higher CCN concentrations than the simulation with
672 aqueous chemistry, as more CCN are expected with aqueous chemistry. However, the dry deposition
673 flux (ddmass) also increases in dense fog which causes rapid loss in CCN and activated aerosols
674 during fog events with the AR feedback (Fig. S7) and more so without aqueous-phase chemistry.
675 Shao et al. (2023) examined aerosol-fog interactions for two consecutive fog events by comparing
676 WRF-Chem results with current emissions strengths to those with low emission strengths. They show
677 that the first fog event promotes formation of the second fog event leading to wider fog distribution,
678 and longer fog lifetime favoured by multiple feedbacks including AR feedback i.e., low temperature,
679 high humidity and high stability similar to our study. While Shao et al. (2023) observe a delay in
680 dissipation of the first event and early formation of second fog event, we find an early dissipation and
681 early formation of fog with AR feedback as discussed earlier in this section. In summary, aqueous
682 phase chemistry together with AR feedback promotes early formation of fog while AR feedback alone
683 promotes early dissipation of fog and plays a critical role in the formation and evolution of the fog
684 over IGP.

685

686 8 Conclusions

687 The effects of aerosol-radiation (AR) feedback and aqueous chemistry in air quality and fog
688 have been assessed over IGP. We carried out three experiments using WRF-Chem testing different
689 combinations of PBL schemes and meteorology initial and boundary conditions. The best
690 representation of surface meteorology for the IGP region for the case study (December 20-24, 2017)
691 used ERA-Interim reanalysis to drive the meteorology and ACM2 PBL scheme with soil moisture
692 nudging to ERA-Interim. With this meteorology configuration for WRF-Chem, evaluation of aerosol
693 concentrations with measurements and the impact of aerosols on atmospheric processes during fog
694 were examined. Further, we included trash-burning emissions to represent anthropogenic chloride
695 aerosols in our configuration. Incorporation of trash burning emissions did improve the model

Commented [CB26]: Revised in response to Reviewer 1, Major comment9, and Minor comment 7

Response to Reviewer2, Major comment5

Commented [CB27]: Modified based on Response to Reviewer1, major comment 2

696 simulations of PM_{2.5} and better captured the day-to-day variability of PM_{2.5} in IGP, however
697 underestimated its magnitude compared to CPCB observations. Moreover, secondary aerosols
698 particularly, chloride aerosols are underestimated in the model. This underestimation is likely caused
699 by a low bias in the residential burning emission inventory and a failure of the emission inventory to
700 represent residential sector emissions from the use of trash as cheap heating fuel properly. AOD
701 regional distribution is predicted well by the model for most of the IGP. However, AOD is
702 underestimated over NWIGP likely due to an underestimation of fugitive emissions during wintertime
703 cold spells.

704 The AR interactions showed a significant impact on meteorology and air quality over IGP. A
705 WRF-chem simulation with AR interactions resulted in a lower PBL height by ~50-270 m compared
706 to a simulation without AR interactions leading to accumulation of aerosols and moisture near the
707 surface. Reduced surface shortwave radiation flux and the surface sensible and latent heat fluxes due
708 to aerosol radiative effect suppressed the turbulence resulting in a stable PBL. The shallow PBL
709 further increased surface PM_{2.5} (> 8 µg m⁻³) and RH (2-8%) over IGP and this positive feedback
710 mechanism promoted thickening of fog over IGP. However, an increase in absorbing aerosols in the
711 PBL gave negative feedback, increasing the shortwave heating and temperature particularly over
712 CIGP. Fog forms when air is saturated which occurs when the surface temperature is reduced or the
713 moisture content increases causing saturation of air. This study suggests that increase in RH saturated
714 the air and the increase in aerosols favoured fog formation as depicted by the thickening of fog
715 intensity. Aqueous phase chemistry on the other hand contributed significantly to secondary aerosols
716 in the fog, especially sulfate aerosols, indicating substantial formation of secondary aerosols in the
717 cloud. The underpredicted secondary aerosols over NWIGP where no fog occurred implies
718 underestimation of formation of aerosols through gas and aerosol chemistry in the model. This
719 underestimation could also be linked to an underestimation of pH in the default MOSAIC scheme
720 (Ruan et al., 2022) which slows the secondary aerosol formation, or an underestimation of the aqueous
721 sulfur oxidation in haze aerosol at > 80% RH before the onset of fog (Acharja et al., 2022), or missing
722 multiphase oxidation processes (Wang et al., 2022). Nevertheless, we find that the model successfully
723 simulates the same changes in the inorganic composition during fog in IGP as reported by
724 observational studies referred earlier in section 6. We also observed that AR feedback with aqueous
725 chemistry initiated the fog formation 1-2 hours earlier than the initiation time in the simulation
726 without AR feedback and without aqueous phase chemistry whereas AR feedback alone led to early
727 dissipation of fog. In addition, fog acted as an important sink of aerosols in a polluted environment
728 with increased dry deposition with cloud water. Thus, AR feedback and aqueous chemistry play a
729 significant role in modulating the distribution and concentration of aerosols and evolution of fog in the
730 PBL.

731 The large emission of aerosols and trace gases in the IGP makes the atmospheric dynamics as
732 well as chemistry complex, suggesting the need for more studies using both models and ground-based

733 measurements to better understand the processes. While all aerosol types interact with solar radiation
734 and reduce the surface reaching flux, presence of absorbing aerosols in the boundary layer and its
735 vertical distribution plays an important role in modulating the meteorology over IGP. It is therefore
736 crucial to improve the simulation of absorbing aerosols e.g., BC in the vertical as well as at the surface
737 to increase the accuracy in predicting formation as well as the dissipation of fog in this region.
738 Emissions from burning for residential heating are an important source of aerosols in IGP during post-
739 monsoon and winter and the inclusion of these sources in the emission inventory would improve the
740 prediction of wintertime aerosols. For example, the underestimation of chloride aerosol in the model
741 indicates unaccounted emission sources over IGP and the need for more work on better quantifying
742 trash burning emissions, which may not only improve particulate chloride in the model but also
743 improve simulations of other aerosol chemical components through aerosol thermodynamics.
744 Additionally, more detailed modeling studies are required to understand the missing chemical
745 processes if any in the model which leads to biases in sulfate, nitrate and ammonium partitioning
746 between gas and aerosol phases. We find that the change in PBL height with AR feedback is sensitive
747 to changes in LH, signifying the role of soil moisture in PBL dynamics. Several studies have reported
748 cooling over IGP due to an increase in irrigation (Kumar et al., 2017; Mishra et al., 2020). Further
749 investigations into the role of irrigation in the increasing fog events over NWIGP would help in better
750 understanding the formation and persistence of fog over this region. It can be concluded that fog
751 forecasting is a complex process due to the multiple factors involved and this work suggests that AR
752 feedback is important in fog forecasting while aqueous phase chemistry plays an important role in
753 defining the composition of aerosols over IGP.

754

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770

771 **Data availability:** All the model simulations are archived on the NCAR campaign storage
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773 corresponding author. WIFEX data can be made available by contacting Dr S.D. Ghude. Trash
774 Burning emission data is available on Mendeley data (doi- <http://dx.doi.org/10.17632/t2tn4t9473.1>).
775 MODIS AOD retrievals can be downloaded from <https://earthdata.nasa.gov/>.

776

777 **Author contributions:**

778 CB: Conceptualization, Formal Analysis, Writing

779 MB: Conceptualization, Supervision, Writing-review and editing, Funding acquisition

780 RK: Conceptualization, Supervision, Writing-review and editing

781 SDG: provided ground-based observation data, Writing-review and editing

782 VS and BS: provided trash burning emission data, Writing-review and editing

783

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785

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1203 Table 1 Experiment set-up for the study. Numbers in parentheses for the physics options denote the
 1204 namelist settings of the WRF-Chem model.

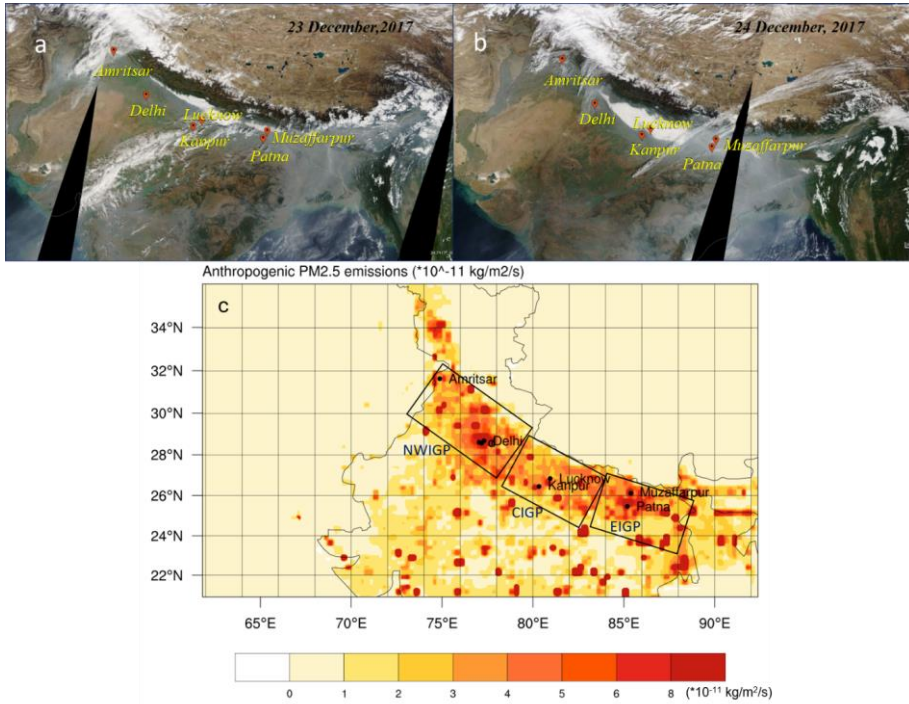
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	EXP 1	EXP 2	EXP 3
Meteorology Initial /lateral boundary Condition:	<i>NCEP Final Analysis (GFS-FNL), 1° x 1°, 6 hourly</i>	<i>ERA-Interim Project, 1.125° x 0.703°, 6 hourly</i>	<i>ERA-Interim Project, 1.125° x 0.703°, 6 hourly</i>
Physics Options			
Cloud Physics	<i>Morrison 2-mom (10)</i>	<i>Morrison 2- mom (10)</i>	<i>Morrison 2- mom (10)</i>
Longwave Radiation	<i>RRTMG scheme (4)</i>	<i>RRTMG scheme (4)</i>	<i>RRTMG scheme (4)</i>
Shortwave Radiation	<i>Goddard shortwave (2)</i>	<i>RRTMG scheme (4)</i>	<i>RRTMG scheme (4)</i>
Surface Layer Physics	<i>Revised MM5 Monin-Obukhov scheme (1)</i>	<i>Revised MM5 Monin-Obukhov scheme (1)</i>	<i>Pleim-Xiu (7)</i>
Surface Model	<i>unified Noah land-surface model (2)</i>	<i>NoahMP (4)</i>	<i>Pleim-Xiu (7)</i>
PBL Scheme	<i>YSU scheme (1)</i>	<i>YSU (1)</i>	<i>ACM2 (7)</i>
Convective Parameterization	<i>Grell-Freitas (3)</i>	<i>Grell-Freitas (3)</i>	<i>Grell-Freitas (3)</i>
	<i>Continuous simulation</i>	<i>*Meteorology refreshed every 24 hr</i>	<i>**Continuous simulation: Soil nudging included</i>

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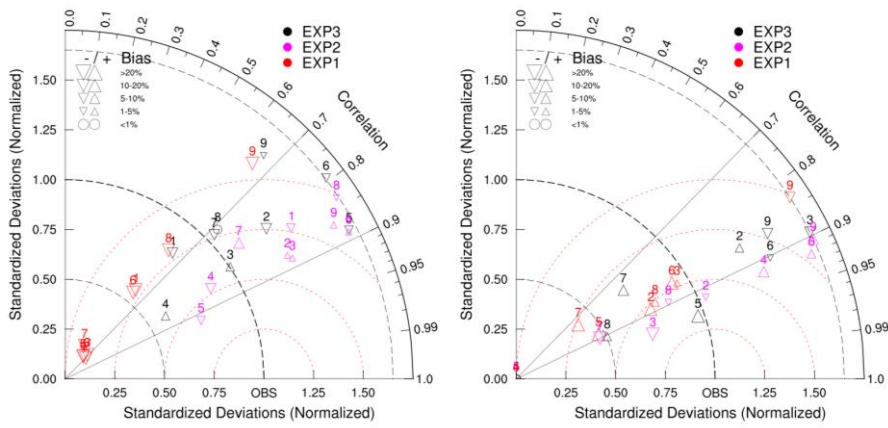
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 1210 Figure 1 The MODIS reflectance (true color) map representing low cloud over Indo Gangetic Plains,
 1211 India (study region) indicative of likely fog and haze on 23rd December (a) and 24th December
 1212 (b)2017. (c) Anthropogenic emission of PM_{2.5} over IGP for December 2017 obtained from EDGAR-
 1213 HTAP. The boxes represent the regions Northwest IGP (NWIGP), Central IGP (CIGP), and East IGP
 1214 (EIGP).

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Commented [CB28]: Modified in response to Reviewer1, Minor comment5

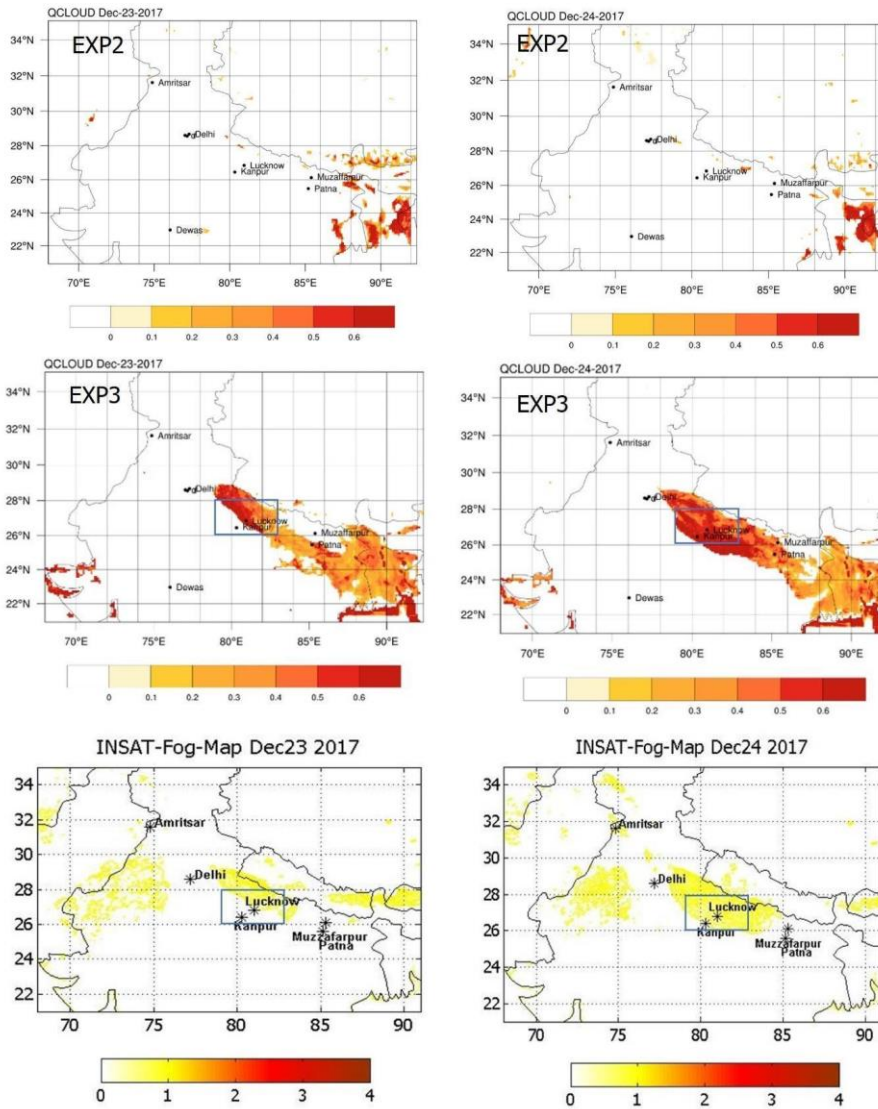


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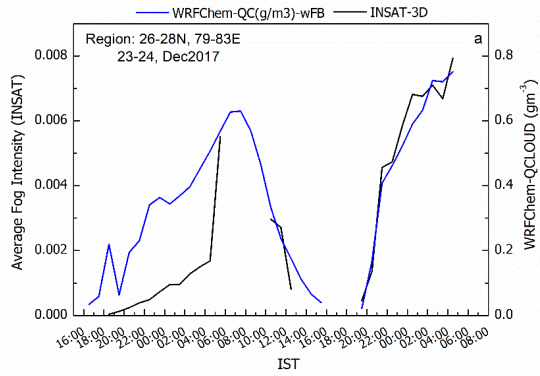
1219 **Figure 2** Taylor Diagram of simulated (WRF-Chem) and observed (CPCB) relative humidity (left) and
 1220 2-m temperature (right) over IGP. The colors indicate the experiments. The red dotted contours
 1221 represent RMS values. The marker (triangles) size varies with a mean bias between the experiments
 1222 and observation. Upside-down triangles represent positive bias (exp-obs) and vice versa. The stations
 1223 over IGP are denoted by number 1. Amritsar, 2. IGI Airport (Delhi), 3. IHBAS (Delhi), 4. Dwarka
 1224 (Delhi), 5. RKP (Delhi), 6. Kanpur, 7. Lucknow, 8. Patna, 9. Muzaffarpur. The locations are marked
 1225 in Fig.1a.

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Commented [CB29]: Modified according to reviewer1, minor comment 8



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 1228 Figure 3 Comparison of fog coverage from WRF-Chem and INSAT-3D satellite for 23 and 24 Dec
 1229 2017. WRF-Chem fog is represented by surface layer cloud water mixing ratios (in g m^{-3}) whereas
 1230 INSAT-3D provides fog intensity which varies from 0 to 4 indicating SHALLOW, MODERATE,
 1231 DENSE, and VERY_DENSE, respectively. The rectangle in central IGP is the region for the time
 1232 series analysis.
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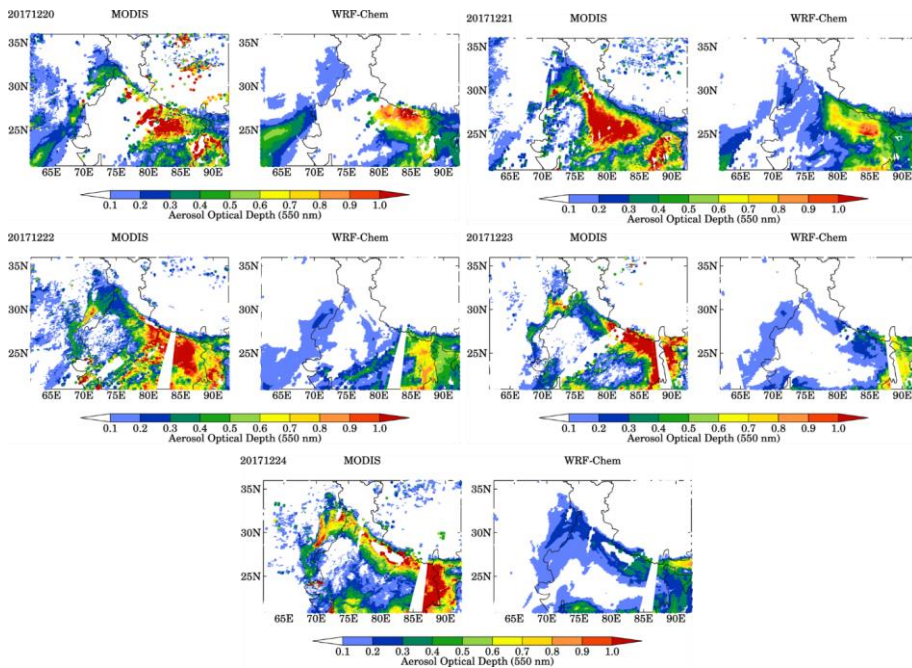


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1235 Figure 4 Average Hourly variation of fog on 23 and 24 December 2017 from WRF-Chem EXP3
 1236 simulation and INSAT-3D satellite between 26°N-28°N,79°E-83°E (region shown in Fig 3). The time
 1237 is in IST (Indian Standard Time; IST is 5.5 hours ahead of Universal Time Coordinate (UTC).

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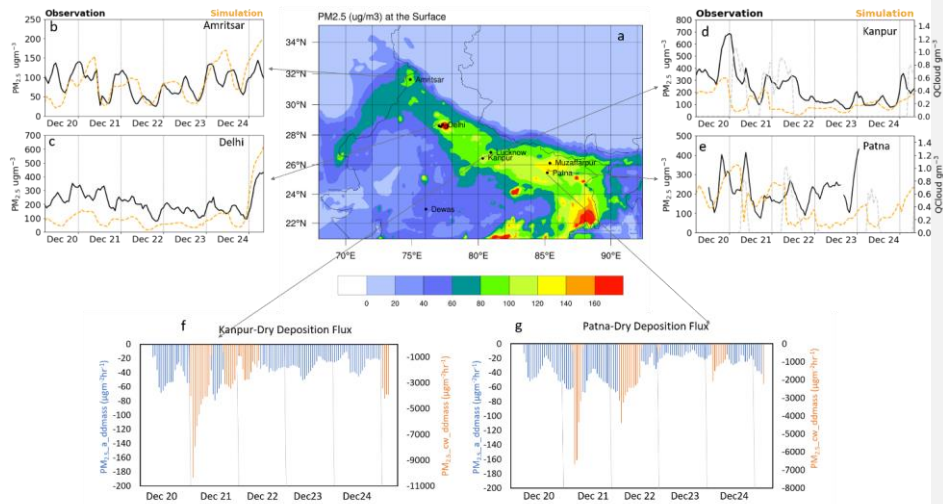
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1241 Figure 5 Comparison of WRF-Chem AOD with MODIS observation over the model domain on 20,
 1242 21, 22, 23, and 24 December 2017.

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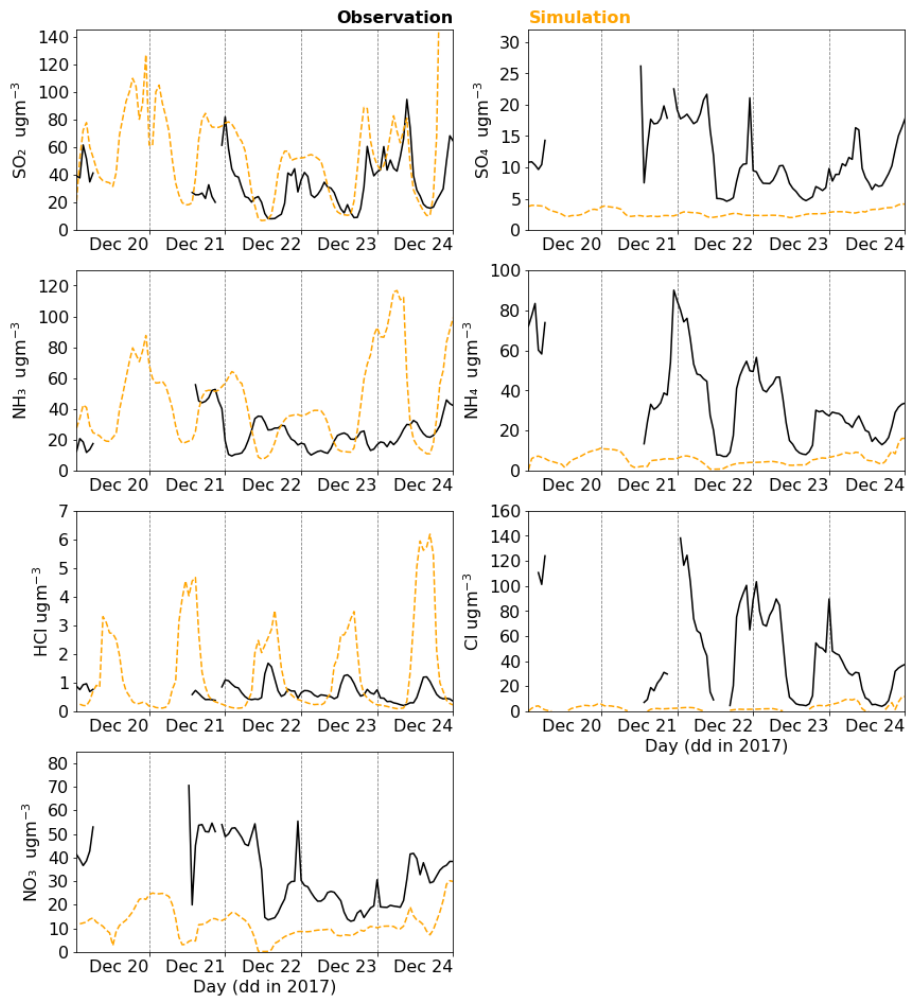
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1247 Figure 6 WRF-Chem simulated surface PM_{2.5} map over IGP (a); comparison of WRF-Chem PM_{2.5}
1248 with CPCB observation for the period 20-24 Dec 2017 for (b) Amritsar, (c) Delhi, (d) Kanpur and (e)
1249 Patna. Dry Deposition rate of PM_{2.5} for (f) Kanpur and (g) Patna. The grey dotted line in (d) Kanpur
1250 and (e) Patna is fog (QCloud) present during the study period.

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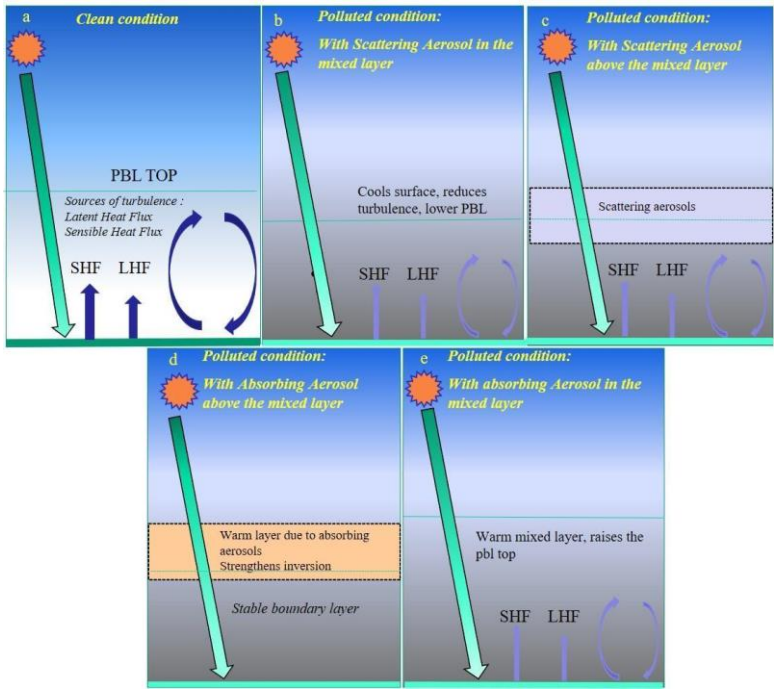


1252 Figure 7 Comparison of WRF-Chem simulated ions (SO_4^{2-} , NH_4^+ , NO_3^- , Cl^-) and trace gases (SO_2 ,
 1253 NH_3 & HCl) with the observation from WIFEX campaign at Delhi.
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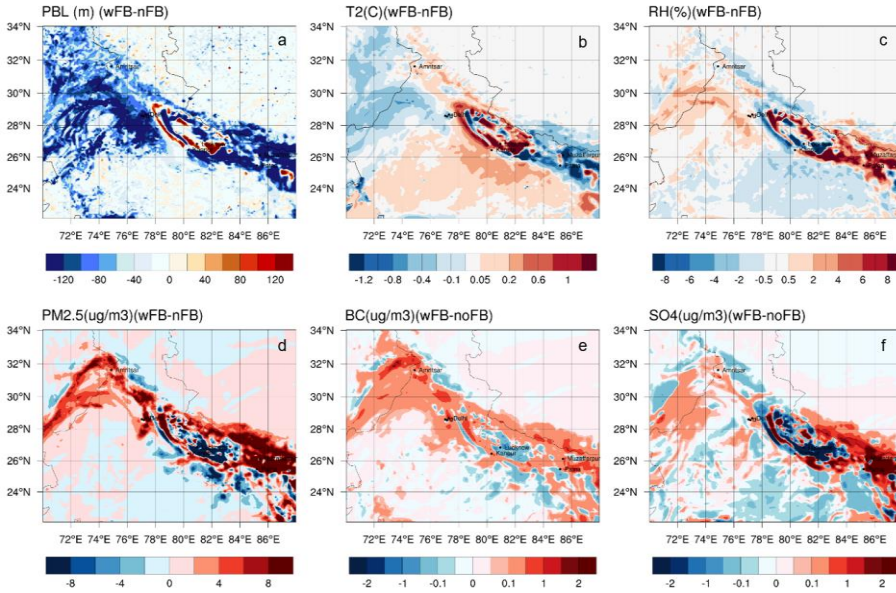


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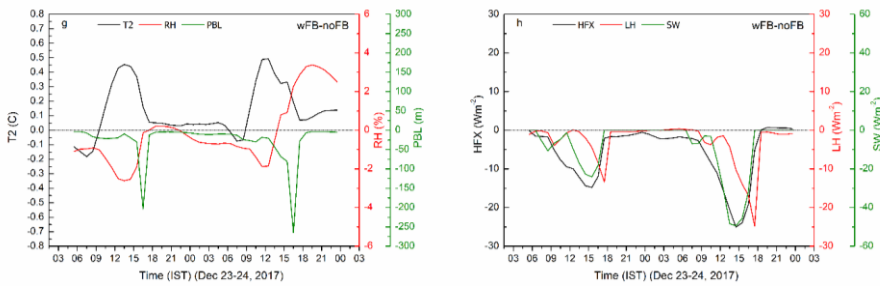
1259 Figure 8 Schematic diagram of Aerosol Radiation Feedback.

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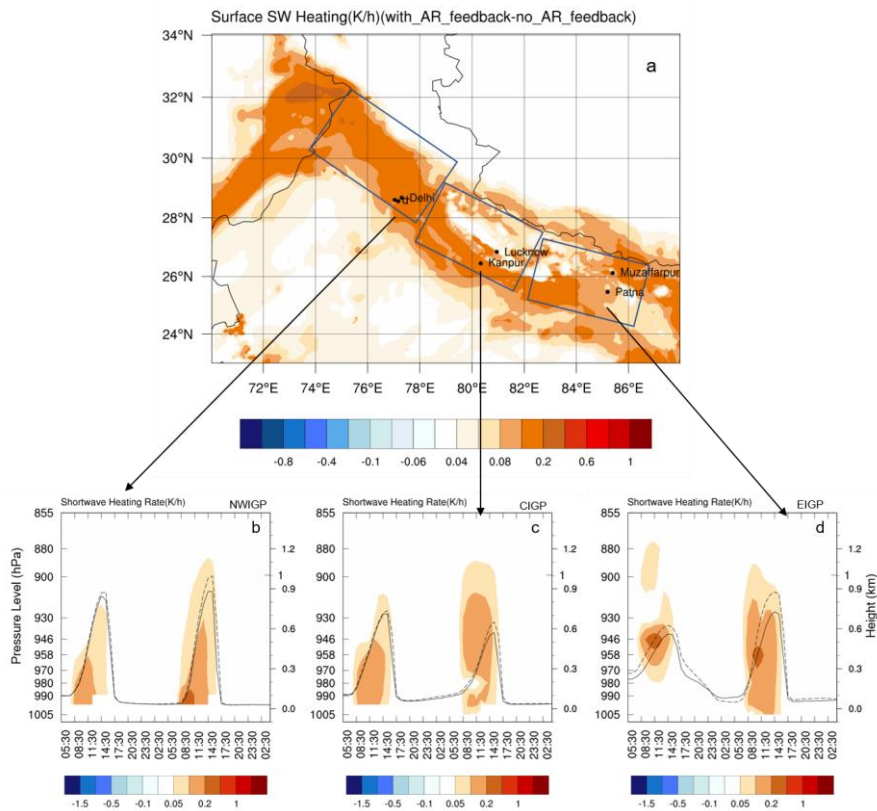
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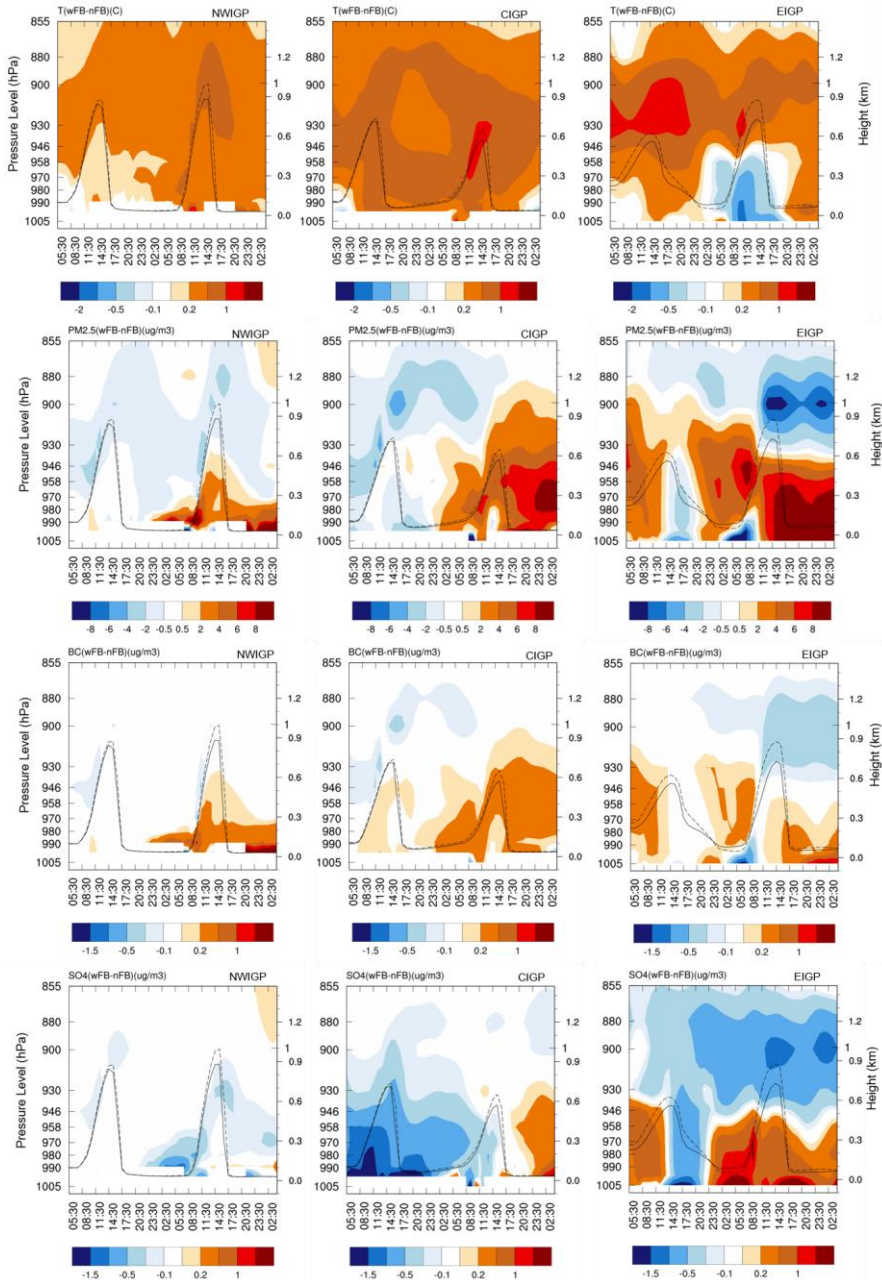
1264 Figure 9 Effect of Aerosol Radiation feedback (wFB-nFB) on (a) PBL height, (b) 2-m temperature, (c)
 1265 2-m relative humidity, (d) surface $PM_{2.5}$, (e) surface BC and (f) surface SO_4 for December 24 at local
 1266 noon (13:30-15:30 IST). (g) The time series of ΔPBL , ΔT_2 , and ΔRH ; (h) ΔHFX (sensible heat flux),
 1267 ΔLH (latent heat flux), and ΔSWF (downward shortwave flux) over CIGP for December 23 and 24. Δ
 1268 denotes the difference between with and without AR feedback (wFB-nFB).

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1272 Figure 10 Differences in shortwave heating rates (K h⁻¹) between simulations with and without aerosol
1273 radiation feedback (a) at the surface, and for pressure-time cross-sections over (b) NWIGP, (c) CIGP.
1274 And (d) EIGP for December 23 and 24. The solid and dashed lines are the PBL height with and
1275 without AR feedback respectively. The time is in IST.

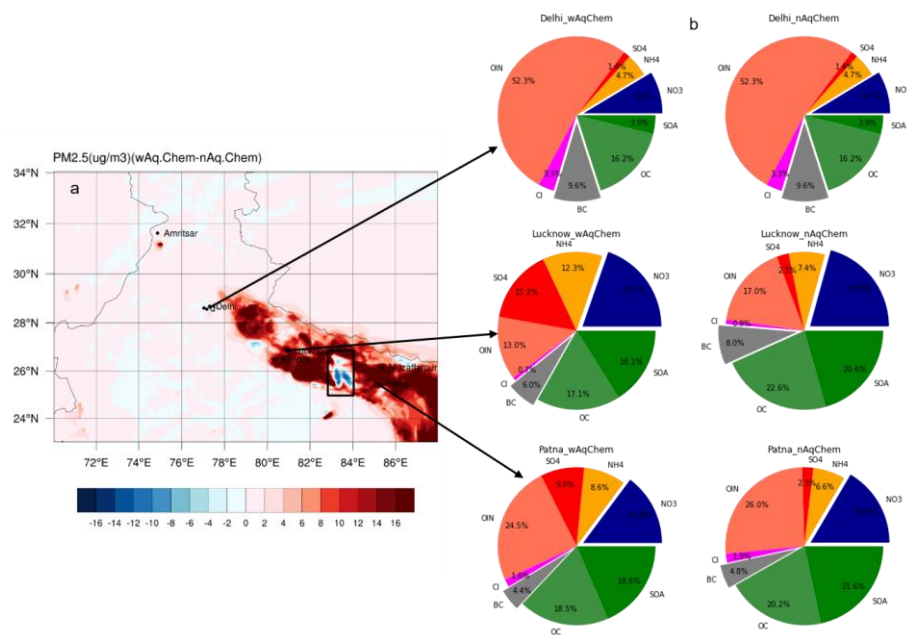


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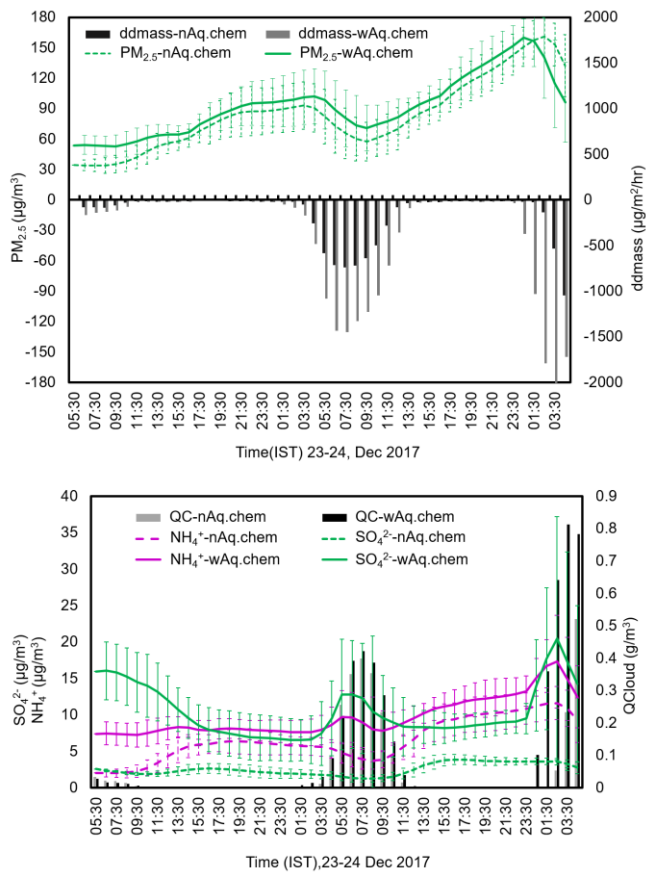
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1279 Figure 11 Pressure-time cross-section of the differences in T, PM_{2.5}, BC and SO₄²⁻ between
 1280 simulations with and without the AR feedback for December 23 and 24. The solid and dashed lines
 1281 are the PBL height with and without AR feedback respectively. The time is in IST.
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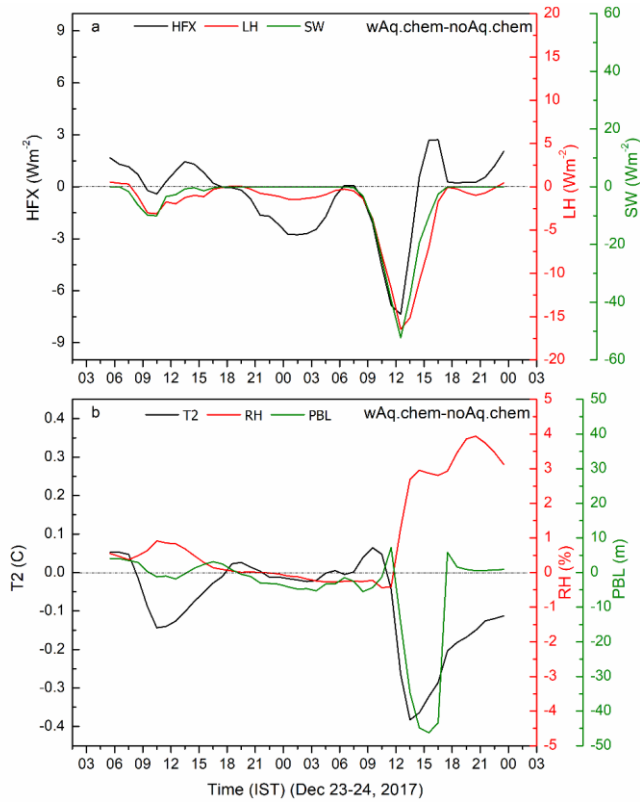
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 1285 Figure 12 (a) Surface $\Delta PM_{2.5}$ (wAq.chem-noAq.chem) and (b) pie charts of PM_{2.5} composition
 1286 distribution for the two cases, with and without Aqueous phase Chemistry for 24 Dec 2017. The
 1287 stations Delhi, Lucknow (LKN), and Patna are representative of NWIGP, CIGP, and EIGP regions
 1288 respectively.
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 1293 **Figure 13** Time series of (a) PM_{2.5} and its dry deposition (ddmass) flux change, (b) SO₄²⁻, NH₄⁺ and
 1294 LWC (QCloud) with and without aqueous phase chemistry included in the model, averaged over the
 1295 region bounded by a black rectangle in Fig. 12, for 23 and 24 December, 2017.

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Commented [CB30]: Standard deviation added in response to Reviewer1, Major comment6
Commented [CB31]: ddmass expanded response to Reviewer 1, Minor comment 9



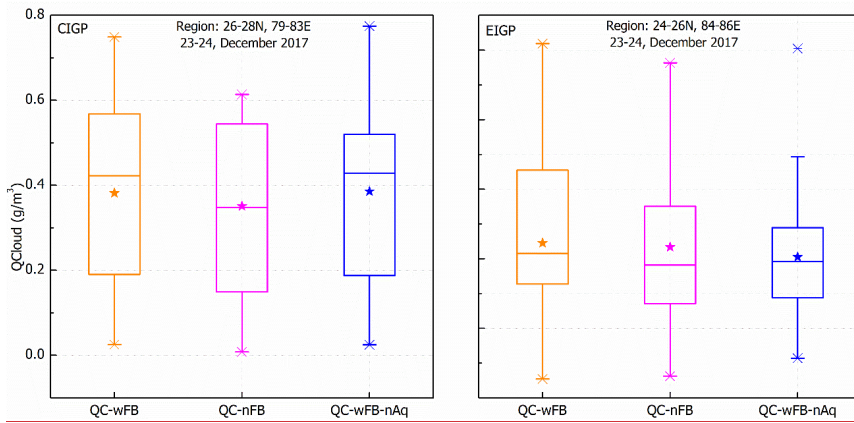
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1300 Figure 14 Time series of (a) Δ HFX (sensible heat flux), Δ LH (latent heat flux), and Δ SWF
 1301 (downward shortwave flux); (b) Δ T2, Δ RH, and Δ PBL over CIGP (79E-83E,26N-28N), for
 1302 23 and 24 December, 2017. Δ denotes the difference between with and without aqueous phase
 1303 chemistry.

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1308 Figure 15 Averages (stars), medians (horizontal lines), quartiles (boxes), maxima, and minima for
 1309 LWC (QCloud) averaged over CIGP (left panel) and EIGP (right panel) for the fog event on 23-24
 1310 December 2017. Gold is for the simulation with AR feedback and aqueous chemistry, magenta for the
 1311 simulation with no AR feedback but includes aqueous chemistry, and blue for the simulation with AR
 1312 feedback but no aqueous chemistry. WRF-Chem does not produce fog in the NWIGP during the study
 1313 period.

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Table 2: Table showing the start and end time of fog1 on 23-24 December 2017 with LWC for the sensitivity experiments, with AR feedback, no AR feedback and no Aqueous phase chemistry

Fig 1 (December 23-24, 2017)

	EXP-wFB			EXP-nFB			EXP-nAq.Chem		
	Start time (IST)	End time (IST)	Duration of Fog	Start time (IST)	End time (IST)	Duration of Fog	Start time (IST)	End time (IST)	Duration of Fog
CI GP	16:30	15:30	23h	18:30	17:30	23h	18:30	15:30	21h
LWC (g/m ³)	0.034	0.036±0.032		0.141±0.154	0.068±0.005		0.184±0.138	0.034 ±0.021	
Kanpur	05:30	13:30	8h	05:30	12:30	7h	05:30	12:30	7h
LWC (g/m ³)	0.334±0.487	0.017		0.458±0.357	0.173±0.071		0.533	0.025±0.0123	
Lucknow	23:30	14:30	15h	00:30	14:30	14h	23:30	14:30	15h
LWC (g/m ³)	0.269±0.145	0.087±0.040		0.232±0.132	0.029±0.024		0.139±0.084	0.025±0.012	
EIGP	21:30	12:30	15h	23:30	10:30	11h	21:30	10:30	13h
LWC (g/m ³)	0.099±0.092	0.007		0.198±0.188	0.084±0.060		0.026±0.008	0.153±0.119	
Patna	00:30	12:30	12h	04:30	10:30	6h	02:30	10:30	8h
LWC (g/m ³)	0.100±0.090	0.007		0.009±0.005	0.038±0.041		0.196±0.198	0.166±0.130	
Muzzafarpur	05:30	11:30	6h	06:30	10:30	4h	06:30	09:30	3h
LWC (g/m ³)	0.112±0.146	0.043±0.057		0.051±0.041	0.003		0.142±0.151	0.157±0.064	

Commented [CB32]: Table added in response to Reviewer1, Minor comment7

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1332 **Table 2:** Table showing the start time of fog 2 on 24 December 2017 with LWC for the sensitivity
1333 experiments, with AR feedback, no AR feedback and no Aqueous phase chemistry. Fog2 end time
1334 could not be noted as simulation ended on 25 December 2017, 00UT (5:30 IST) before fog2
1335 dissipates.

Commented [CB33]: Table added in response to Reviewer1, Minor comment7

Fog 2 (December 24, 2017)			
Start time			
(IST)			
	EXP-wFB	EXP-nFB	EXP-nAq.Chem
CIGP	19:30	20:30	21:30
LWC (g/m ³)	0.025	0.008±0.007	0.025
Kanpur	21:30	22:30	23:30
LWC (g/m ³)	0.041±0.007	0.298±0.218	0.482±0.398
Lucknow	21:30	20:30	00:30
LWC (g/m ³)	0.203±0.165	0.005	0.229±0.209
EIGP	00:30	01:30	01:30
LWC (g/m ³)	0.024±0.030	0.072±0.088	0.014±0.009
Patna	03:30	03:30	03:30
LWC (g/m ³)	0.030± 0.046	0.018	0.060
Muzzafarpur	04:30	No fog	No fog
LWC (g/m ³)	0.159±0.038		

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