



1 **Effects of spatial resolution on WRF v3.8.1 simulated meteorology over the central**

2 **Himalaya**

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15

16 **Abstract**

17 The sensitive and fragile ecosystem of the central Himalayan (CH) region, experiencing enhanced
18 anthropogenic pressure, requires adequate atmospheric observations and an improved representation of
19 Himalaya in the models. However, the accuracies of atmospheric models remain limited here due to highly
20 complex mountainous topography. This article delineates the effects of spatial resolution on the modeled
21 meteorology and dynamics over the CH by combining the WRF (Weather Research and Forecasting)
22 model with the GVAX (Ganges Valley Aerosol Experiment) observations during the summer monsoon.
23 WRF simulation is performed over a domain (d01) encompassing northern India at 15 km x 15 km
24 resolution, and two nests: d02 (5 km x 5 km) and d03 (1 km x 1 km) centered over CH with boundary



25 conditions from respective parent domains. WRF simulations reveal higher variability in meteorology e.g.
26 Relative Humidity (RH=71.4–93.3%), Wind speed (WS=1.6–3.1 ms⁻¹), as compared to the ERA Interim
27 reanalysis (RH=79.4–85.0, and WS=1.3–2.3ms⁻¹) over the northern India owing to higher resolution. WRF
28 simulated temporal evolution of meteorological profiles is seen to be in agreement with the balloon-borne
29 measurements with stronger correlations aloft ($r = 0.44$ – 0.92), than those in the lower troposphere ($r =$
30 0.27 – 0.48). However, the model overestimates temperature (warm bias by 2.8°C) and underestimates RH
31 (dry bias by 7.6%) at surface in the d01. Model results show a significant improvement in d03 (P=827.6
32 hPa, T=19.8°C, RH=90.2%) and are closer to the GVAX observations (P=801.3, T=19.5, RH=94.5%).
33 Temporal variations in near surface P, T and RH are also reproduced by WRF d03 to an extent ($r > 0.5$).
34 A sensitivity simulation incorporating the feedback from nested domain demonstrated improvements in
35 simulated P, T and RH over CH. Our study shows the WRF model set up at finer spatial resolution can
36 significantly reduce the biases in simulated meteorology and such an improved representation of CH can
37 be adopted through domain feedback into regional-scale simulations. Interestingly, WRF simulates a
38 dominant easterly wind component at 1 km x 1 km resolution (d03), which was missing in the coarse
39 simulations; however, a frequent southeastward wind component remained underestimated. Model
40 simulation implementing a high resolution (3 s) topography input (SRTM) improved the prediction of
41 wind directions, nevertheless, further improvements are required to better reproduce the observed local-
42 scale dynamics over the CH.

43 1. Introduction

44 Himalayan region is one of the most complex and fragile geographical systems in the world, and has
45 paramount importance for the climatic implications and air composition at regional to global scales (e.g.
46 Lawrence et al., 2010, Pant et al., 2018; Lelieveld et al., 2018). The ground-based observations of
47 meteorology and fine-scale dynamics are highly sparse. In this direction, an intensive field campaign
48 called as the GVAX (Kotamarthi, 2013) was carried out over a mountainous site in the Central Himalaya



49 which provided valuable meteorological observations for atmospheric research, model evaluation and
50 improvements. Accurate simulations of meteorology are needed for numerous investigations, such as to
51 study the regional and global climate change, snow-cover change, trapping and transport of regional
52 pollution, and the hydrological cycle especially monsoon system (e. g. Sharma and Ganju, 2000; Bhutiyani
53 et al., 2007; Pant et al., 2018). Studies focussing over this region have become more important due to the
54 increasing anthropogenic influences resulting in enhanced levels of Short-Lived Climate forcing Pollutants
55 (SLCPs) along the Himalayan foothills (e. g. Ojha et al., 2012; Sarangi et al., 2014; Rupakheti et al., 2017;
56 Deep et al., 2019; Ojha et al., 2019). Although Global Climate Models (GCMs) simulate the climate
57 variabilities over global scale, their application for reproducing observations in the regions of complex
58 landscapes is limited, due to coarse horizontal resolution (e. g. Wilby et al., 1999; Boyle et al. 2010;
59 Tselioudis et al., 2012; Pervez and Henebry, 2014; Meher et al., 2017). Mountain ridges, rapidly changing
60 land-cover, and the low altitude valleys often lie within a grid box of typical global climate models
61 resulting in significant biases in model results when compared with observations (e. g. Ojha et al., 2012;
62 Tiwari et al., 2017, Pant et al., 2018). On the other hand, Regional Climate Models (RCMs) at finer
63 resolutions allow better representation of the topographical features thus providing improved simulations
64 of the atmospheric variability over regions of complex terrains. Several mesoscale models (e. g.
65 Christensen et al., 1996; Caya and Laprise 1999; Skamarock et al., 2008; Zadra et al., 2008) have been
66 developed and applied successfully over different parts of the world. These studies have revealed that the
67 RCMs provide significant new insights by parameterizing or explicitly simulating atmospheric processes
68 over finer spatial scales. Nevertheless, large uncertainties are still seen over highly complex areas
69 indicating the effects of further unresolved terrain features (e. g. Wang et al., 2004; Laprise, 2008; Foley,
70 2010) and need to improve the simulations.

71 Of late, anthropogenic influences and climate forcing have been increasing over the Himalaya and its
72 foothill regions since pre-industrial times (Pant et al., 2006; Bonasoni et al., 2012; Srivastava et al., 2015).
73 Further, an increase in the intensity and frequency of extreme weather events have been observed over the



74 Himalayan region (e. g. Nandargi and Dhar, 2012; Sun et al., 2017; Dimri et al., 2017) in past few decades.
75 These events include extreme rainfall and resulting flash floods, cloudbursts, landslides etc., and their
76 causes range from mesoscale processes to larger synoptic scale events. Unfortunately, the lack of
77 observational network covering the Himalaya and foothills with sufficient spatio-temporal density inhibits
78 the detailed understanding of the aforementioned processes, and meteorological and dynamical conditions
79 in the region. Therefore, usage of regional models, evaluated against available in-situ measurements would
80 fill the gap of investigating atmospheric variability in the observationally sparse and geographically
81 complex terrain of Himalaya.

82 In this regard, a few studies have applied the regional model WRF to simulate the variations in the
83 meteorology, winds, and boundary layer dynamics over the Himalaya and foothills (Kumar et al., 2012;
84 Sarangi et al., 2014; Singh et al., 2016; Mues et al., 2018). WRF model with suitably chosen schemes is
85 generally able to reproduce the spatio-temporal variations in the regional-scale meteorology and wind
86 patterns (Kumar et al., 2012) and to an extent also captured the mountain-valley wind systems (Sarangi et
87 al., 2014) and boundary layer dynamics (Singh et al., 2016; Mues et al., 2018). However, most of these
88 studies utilized model at horizontal resolutions of 45 to 30 km, except study by Singh et al. (2016) used
89 relatively higher spatial resolution (5 km x 5 km). Nevertheless, it remains unclear how the finer resolution
90 could better resolve the complex terrain of the central Himalaya and improve the meteorological
91 simulations, especially at 5 km to 1 km resolution.

92 With this opportunity of model evaluation and improvements in simulating meteorological and dynamical
93 variability over the CH, here, we have used a nested WRF set up with a coarse 15 km x 15 km domain
94 (d01) with nests of 5 km x 5 km (d02) and 1 km x 1 km (d03) centred over the CH. The main objectives
95 of the study are as follows:

- 96 1. To examine the model performance over the CH at 15 km x 15 km resolution



- 97 2. To examine the effects or improvements that can be achieved at higher spatial resolutions: 5 km x
98 5 km, and 1 km x 1 km.
- 99 3. To investigate the effect of feedback from nest that could be adopted into parent domain, as this
100 would allow configuring a setup covering larger Indian region with more accurate results over
101 Himalaya
- 102 4. To implement a very high resolution (3 s) topographical input into the model to examine the
103 potential of simulations finer than 1 km in reproducing local-scale dynamics

104 Subsequent section 2 describes the model set up, followed by experimental design, and a discussion of
105 datasets used for model evaluation. Section 3 provides comparison of model results with the ERA Interim
106 reanalysis (section 3.1), radiosonde observations (section 3.2), and ground-based measurements (section
107 3.3). Analysis of domain feedback is presented in section 3.4, and the effect of implementing high
108 resolution topography is investigated in section 3.5, followed by the summary and conclusions in the
109 section 4.

110 **2. Methodology**

111 **2.1 Model set up and Experimental Design**

112 Weather Research and Forecasting (WRF) model–version 3.8.1 has been used in the present study. WRF
113 is a mesoscale non-hydrostatic, Numerical Weather prediction (NWP) model with advance physics and
114 numerical schemes for simulating meteorology and dynamics. WRF-ARW uses Eulerian mass based
115 dynamical core with terrain-following vertical coordinates (Skamarock et al., 2008). ERA Interim
116 reanalysis from the European Center for Medium Range Weather Forecasts (ECMWF) available at a
117 horizontal resolution of $0.7^{\circ} \times 0.7^{\circ}$ at 6 h interval has been used to provide the initial and lateral boundary
118 conditions to the WRF model. Static geographical data, which includes the information of terrain height,
119 land use, and land cover etc., is based on the Moderate Resolution Imaging Spectroradiometer (MODIS)
120 data available at 30s horizontal resolution.

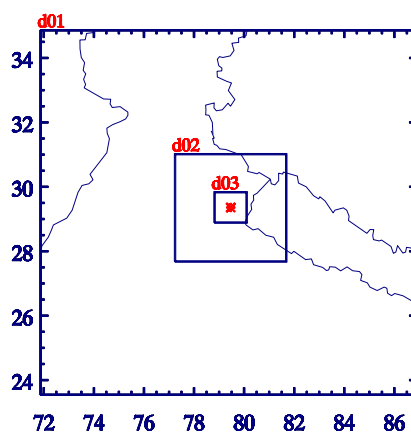


121 The shortwave radiation scheme used is the Goddard scheme (Chou and Suarez, 1994) while the long
122 wave radiation is simulated by the Rapid Radiative Transfer Model (Mlawer et al., 1997) scheme. For
123 resolving the boundary layer processes the first order nonlocal closure based Yonsei University (YSU)
124 scheme (Hong et al., 2006) is used including an explicit entrainment layer with the K-profile in an unstable
125 mixed layer. PBL height is determined from the Richardson number (Ri_b) method in this PBL scheme.
126 Convection is parameterized by the Kain-Fritsch (KF) cumulus parameterization (CP) scheme, accounting
127 for sub-grid level processes in the model such as precipitation, latent heat release and vertical redistribution
128 of heat and moisture as a result of convection (Kain, 1990). With increase in model grid resolution to less
129 than 10 km (known as “grey area”), the CP scheme is usually turned off and an explicit microphysics (MP)
130 scheme is needed to resolve cloud and precipitation processes (Weisman et al., 1997). In the present study,
131 the CP scheme is used for d01 while it is turned off for d02 and d03. The Thomson microphysics containing
132 prognostic equations for cloud water, rain water, ice, snow, and graupel mixing ratios, is used (Thompson
133 et al., 2004). Parameterization of surface processes is done with MM5 Monin-Obukhov scheme and
134 Unified Noah land surface model (LSM) (Chen and Dudhia, 2001; Ek et al., 2003; Tewari et al., 2004).
135 The Noah LSM includes a single canopy layer and four soil layers at 0.1, 0.2, 0.6 and 1m within 2m of
136 depth (Ek et al., 2003).

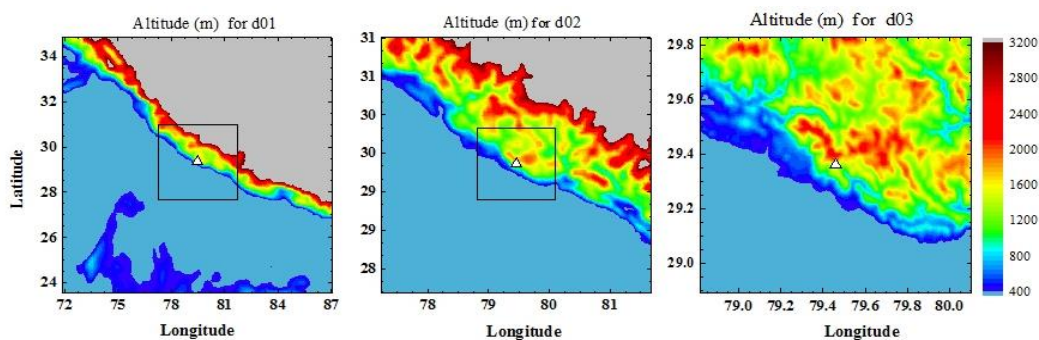
137 The model is configured with three domains of 15 km (d01), 5 km (d02) and 1 km (d03) horizontal grid
138 spacing using Mercator projection centering at Manora Peak (79.46°N, 29.36°E, amsl ~ 1958m) in central
139 Himalaya. Topography within the model domains is highly complex as evident from the ridges (Figure 1).
140 Outer domain d01 includes north part of Thar Desert, part of IGP (Indo-Gangetic-Plain), Himalayan
141 mountains with vegetation and snow cover, while the innermost domain d03 consists of mostly
142 mountainous terrain. Model atmosphere has 51 vertical levels with top at 10 hPa. For d01, 100 east-west
143 and 86 north south grid points are used to account for the effect of synoptic scale meteorology e.g. Indian
144 summer monsoon. The d02 has 88 east-west and 76 north-south grid points covering sufficient spatial
145 region around the observational site to consider the effects of mesoscale dynamics, e.g. change of wind



146 pattern due to orography. The innermost domain 03 has 126 east-west and 106 north-south grid points
147 mainly to reveal local effects e.g. convection, advection, turbulence, orthographic lifting etc.



148



149

150 **Figure 1:** Topography in WRF model domains at three horizontal resolutions: domain d01 (15 x 15km),
151 domain d02 (5 x 5km) and domain d03 (1 x 1 km). Triangle sign indicates the location of the GVAX
152 campaign over Manora Peak, Nainital. Box inside the figure represents the nested domain.

153 The d01 simulation provides the boundary conditions to domain d02, and domain d02 to innermost domain
154 d03. For d01, boundary conditions are provided from ERA Interim reanalysis, as explained earlier. Model
155 simulation has been performed for four months of the summer monsoon season: 01 June 2011 to 30
156 September 2011 (JJAS). First 10 days of the simulation is considered as the spin-up and removed from the
157 analysis. Only the outer domain d01 is nudged with the global reanalysis for temperature, water vapor,



158 zonal and meridional (u and v) components of wind with nudging coefficient of $0.0006 (6 \times 10^{-4})$ at all
159 vertical levels (e.g. Kumar et al., 2012). Several of the configuration options e. g. physics, meteorological
160 nudging, etc. are selected following earlier applications of this model over this region (e. g. Kumar et al.,
161 2012; Ojha et al., 2016; Singh et al., 2016; Sharma et al., 2017).

162 **2.2. Observational data**

163 We utilize the observations conducted as a part of an intensive field campaign- the Ganges Valleys Aerosol
164 Experiment (GVAX) for the evaluation of model simulations. The GVAX campaign was conducted using
165 Atmospheric Radiation Measurement (ARM) Climate Research Facility of the U.S. Department of Energy
166 (DOE) from June, 10 2011 to March 31, 2012 at ARIES, Manora Peak in Nainital (e.g. Kotamarthi, 2013;
167 Singh et al., 2016; Naja et al., 2016). The surface-based meteorological measurements of ambient air
168 temperature, pressure, relative humidity, precipitation, wind (speed and direction) were carried out using
169 an automatic weather station at 1-minute temporal resolution.

170 The vertical profiles of temperature, pressure, relative humidity and horizontal wind (speed and direction)
171 were measured by four launches (00:00, 06:00, 12:00 and 18:00 UTC) of the radiosonde each day during
172 the campaign (Naja et al., 2016). The continuous vertical profiles of the meteorological parameters except
173 wind speed and direction were available from end of June 2011 to the entire period of study, whereas valid
174 wind data were available only for September 2011. Hence, in this study, radiosonde measurements from
175 July 1, 2011 onwards are used for model evaluation of meteorological parameters, except wind speed and
176 direction, which are evaluated for the month of September.

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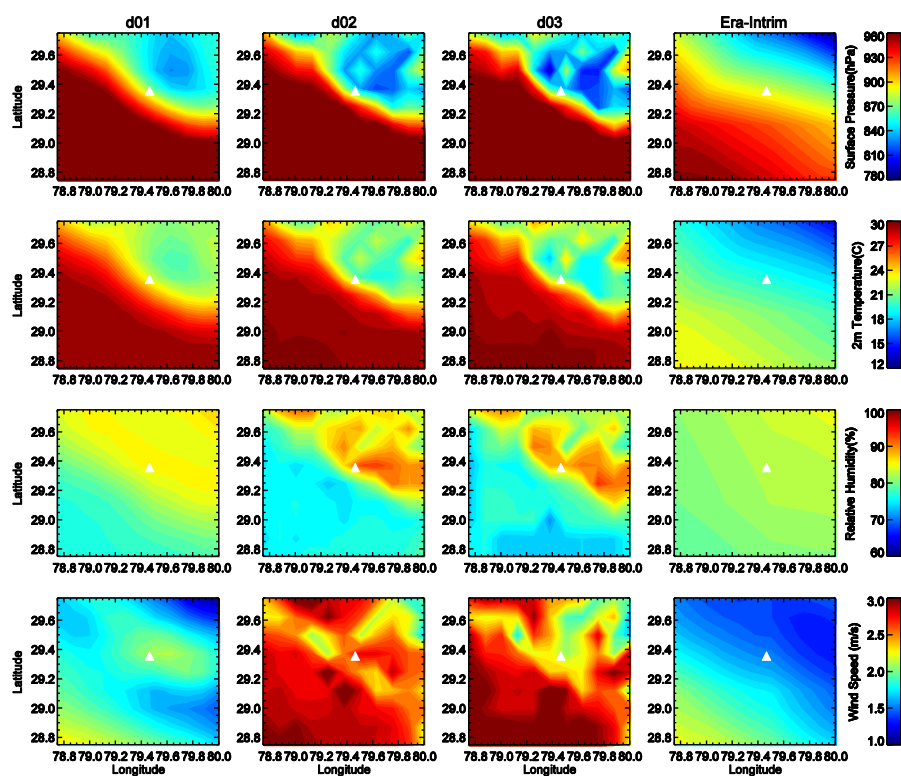
178 **3. Results and Discussions**

179 **3.1. Comparison with ERA Interim reanalysis**

180 ERA Interim reanalysis data set is available globally at resolution of $0.75^{\circ} \times 0.75^{\circ}$ with 60 vertical levels
181 from surface to the top at 0.1 hPa with covering time period from 1979 to present at 6-hourly time step



182 (Dee et. al., 2001). Re-gridded ERA-Interim data are also available at various resolutions such as 0.125,
183 0.25, 0.50, 0.75, 1.0 etc. Here, we have used the highest resolution data available at $0.125^{\circ} \times 0.125^{\circ}$ for
184 comparison with WRF results. We first compare the WRF simulated spatial distribution of meteorological
185 parameters (surface pressure, 2m air Temperature, 2m RH and 10m WS) with ERA Interim reanalysis
186 over common area of all the domains and averaged for the complete simulation period (Figure 2). The
187 common area in all domains includes low-altitude Indo-Gangetic Plain (IGP) region in south (with
188 elevation of less than 400m, Figure 1) and elevated mountains of the central Himalaya in north. Also, for
189 a consistent comparison, model simulated values are taken at the same time intervals as that in ERA
190 Interim data (i.e. every 6h). From the comparison (Figure 2), it is evident that the meteorological
191 parameters simulated by the model are dependent on the model grid resolution. The existence of the sharp
192 gradient topographic height (SGTH) of about 1600 m from the foothill of the Himalaya to the observational
193 site modifies the wind pattern as well as moisture content differently at different grid resolutions,
194 indicating the critical role of mountain orography. The surface pressure explicitly depends upon the
195 elevation of a location from mean sea level. The contour plot of the pressure from ERA Interim shows
196 surface pressure of about 883.9 hPa for observational site Manora Peak, while WRF simulated pressure is
197 839 hPa, 821hPa and 840 hPa for d03, d02 and d01, respectively. WRF simulated surface pressure ranges
198 from 829.0 hPa over high altitude CH region to 979.5 hPa in IGP region within d01. At the same time, the
199 variation range of surface pressure is 817.0–978.8 and 795–977.6 hPa within d02 and d03 respectively,
200 and this is attributed to the improvement in resolved topography on increasing model grid resolution. The
201 effects of the SGTH are not clearly observed for temperature, wind and RH in ERA-Interim contours, and
202 it could be due to the unresolved topographic features. Simulated spatial profiles show significantly
203 distinct meteorology in Indo-Gangetic Plain (IGP), which lies just in the foothills of Himalaya, and
204 elevated central Himalayan region.



205

206 **Figure 2:** Contours in the first three columns show WRF results and the fourth column shows
207 corresponding parameters from the ERA Interim reanalysis. First row shows mean surface pressure during
208 the monsoon (JJAS), second row shows 2m temperature, 3rd row shows 2m relative humidity (RH) and
209 bottom row shows 10 m wind speed.

210 The effect of spatial resolution is clearly observed over the mountainous region of Himalaya, where size
211 of the mountains changes abruptly, with the modelled output showing distinct features with increasing
212 grid resolution. On the other hand, there are minimal differences in the topography of the IGP, and hence
213 features are well captured in the model even at coarser resolution of 15 Km.

214 Model simulations show topography dependent spatial variation in the 2m temperature in the ranges of
215 20.2–29.6°C in d01, 19.2–29.8°C in d02, and 18.0–29.8°C in d03 with lowest values simulated over the



216 elevated mountain peaks and higher values over the temperate IGP region. The contours in three model
217 domains show explicit dependency of 2m temperature on the grid resolution over the mountainous region.
218 With increasing model resolution, the topography is resolved to a greater extent and the lower temperature
219 is simulated at higher surface elevations, as expected. Estimation of water vapour is very important for
220 both climate and numerical weather prediction (NWP) applications. The relative humidity is above 70%
221 in all three domains as the study period is monsoon season. The variations (minimum-maximum) in the
222 humidity in ERA-Interim (80% to 85%) data sets, domains d01 (76-87%), d02 (73 – 93%), and d03 (71-
223 93%) are generally comparable. The mountain slopes provide the uplift to the monsoonal moist air that
224 subsequently saturates on ascent and increases the relative humidity to about 90% as observed over the
225 grid encompassing the site.

226 The wind speed is highly dependent upon the model grid resolution as well as orography-induced
227 circulations during different seasons (Solanki et al., 2016; Solanki et al 2019) as shown in Fig. 2. As
228 mentioned earlier, although the topography of the IGP region does not vary abruptly, the magnitude of the
229 wind speed over this region as well as over the complex Himalayan region are found to change
230 significantly at different model resolutions, thereby indicating that the wind speed is very sensitive to both
231 model resolution and topography. The wind speed in d01 varies from minimum value of 1.1 ms^{-1} to
232 maximum value of 2.3 ms^{-1} and agrees better with the ERA-Interim dataset (1.3 ms^{-1} to 2.3 ms^{-1}), while
233 the wind variations in domains d02 and d03 (1.6 to 3.1 ms^{-1}) are similar but are overestimated as compared
234 to the ERA-interim.

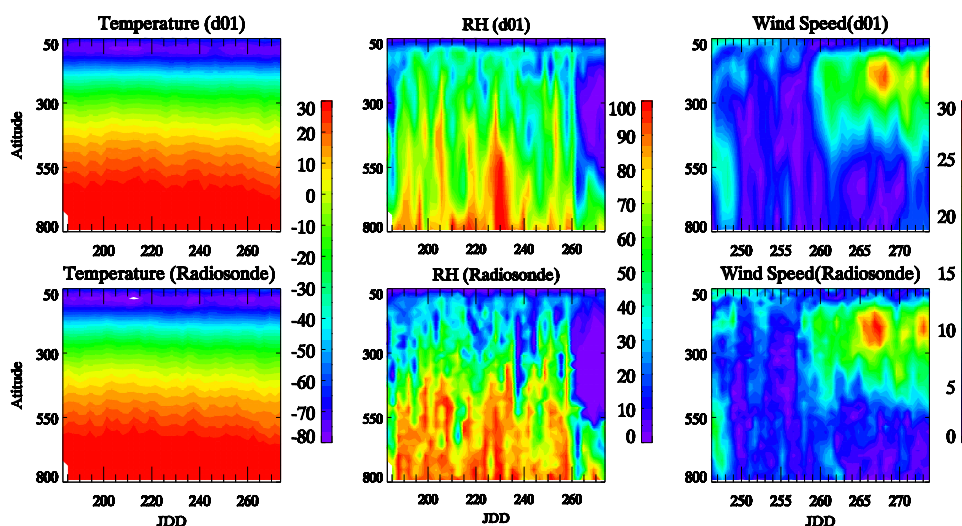
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236 **3.2. Comparison with Radiosonde observations**

237 The vertical profiles of the meteorological parameters: temperature, relative humidity and wind speed are
238 from surface to 50 hPa are shown in the Figure 3 for WRF-d01 simulation and the radiosonde observations.
239 As mentioned earlier, the snapshots of atmospheric profiles are obtained using radiosondes launched four



240 times a day during the GVAX campaign. Various features of the temperature profiles are seen to be well
241 simulated by the WRF model (Figure 3).



242

243 **Figure 3:** The comparison of WRF simulated temperature, relative humidity and wind speed vertical
244 profiles with the radiosonde observations for the pressure-levels from 800 hPa to 50 hPa. Horizontal axis
245 shows the day number of the year 2011 (JDD) starting from 1 July (182th day) to 30 September (273th
246 day). Wind speed profiles are plotted only for month September, 2011.

247

248 The inversion of the temperature at the top of the troposphere occurred at ~100hPa (~16km) in
249 observations, while it occurred at slightly higher altitude (~80hPa) in the model for domain d01. Further
250 WRF model simulates wetter (or more humid) atmosphere at higher altitudes while showing a good
251 agreement with the observations of water content in lower altitudes. The observations of wind speed were
252 available only from 01 September 2011 onwards therefore wind comparison is made only for the
253 September month. The simulated profiles of the wind agree generally well with the observations. For the
254 statistical comparison of the simulated meteorology with the observations, the Taylor diagram (Taylor,
255 2001) is used and shown in Figure 6. In the diagram the comparison is summarized with correlation

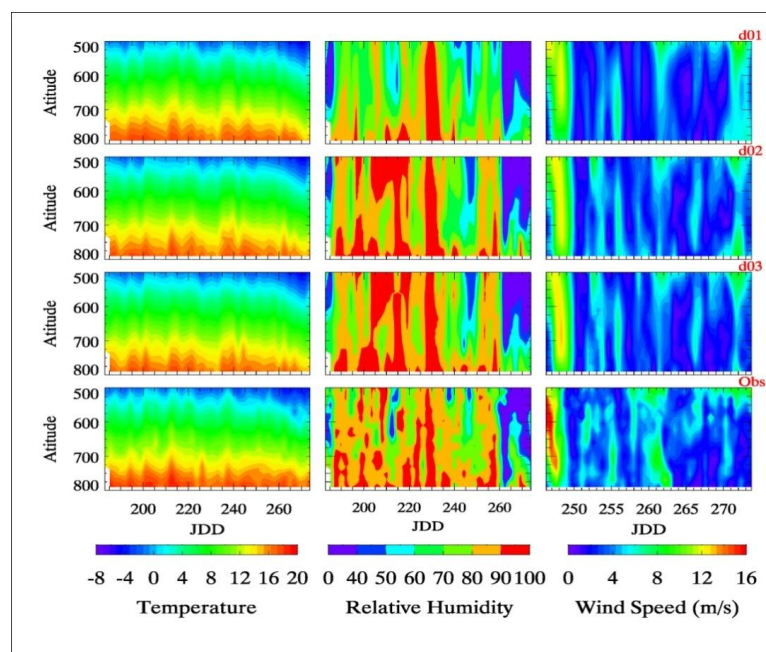


256 coefficient (r), normalized root mean squared difference (RMSD) and normalized standard deviation (SD),
257 normalized to the standard deviation of observation. In most of the cases (shown in Figure 6a) model
258 simulates less variability in meteorological parameters as shown by the normalized standard deviation
259 which turns out to be less than 1. For temperature and wind speed, model shows good agreement at 250hPa
260 ($r > 0.90$) than that in lower altitudes i.e. 750hPa ($r < 0.42$). On the other hand, model captures variability
261 in humidity relatively well at 500 hPa ($r = 0.71$) but shows poor correlation at 50 hPa ($r = 0.17$) near the
262 model top. Lower correlations for temperature and wind speed at 750 hPa pressure near the surface could
263 be due to the terrain induced effects most significant in the local boundary layer. The, surface level winds,
264 and turbulence etc. are some of the features of the boundary layer that are largely affected by the surface
265 and terrain characteristics. The simulated vertical profiles do not show any explicit variation at different
266 grid resolutions in upper troposphere. Therefore a comparison for all three model domains only up to 500
267 hPa is shown in Figure 4.

268

269 Except for the relative humidity in d01, other meteorological parameters shown here do not reveal strong
270 dependencies on the model resolution. The temperature ranges from 20°C at 800 hPa to -8°C at 500 hPa
271 in WRF simulations as well as in the observations. Model however overestimates the relative humidity
272 near 500hPa level in d02 and d03 on some of the days. In case of the wind speed, the model underestimates
273 the magnitude of the wind in first few days up to 500 hPa, though qualitatively the model is able to capture
274 the vertical profiles.

275



276

277 **Figure 4:** A comparison of temperature (first column), relative humidity (second column) and wind speed
278 (third column) profiles for pressure range: 800 to 500 hPa for WRF simulations for all domains (d01: first
279 row, d02: second row, d03: third row) and radiosonde observations (fourth row).

280

281 3.3. Comparison with ground-based observations

282 Model simulated 2m temperature, relative humidity (RH) and 10 m windspeed for the observational site,
283 Manora Peak are compared with the ground-based measurements made during GVAX campaign in Fig.5
284 and summarized in Table1. The diurnal variations in 2m temperature, 2m relative humidity and 10 m wind
285 speed simulated by the WRF model are compared with observations in Figure 5, whereas the surface
286 pressure does not show a significant diurnal variation (not shown here). WRF model simulated 2m
287 temperature shows warm bias in all the three domains. The simulated 2m temperature for d01 varies from
288 20.0 to 24.7°C with the mean value of 22.3°C compared to observed mean value of about 19.5±1.1°C with



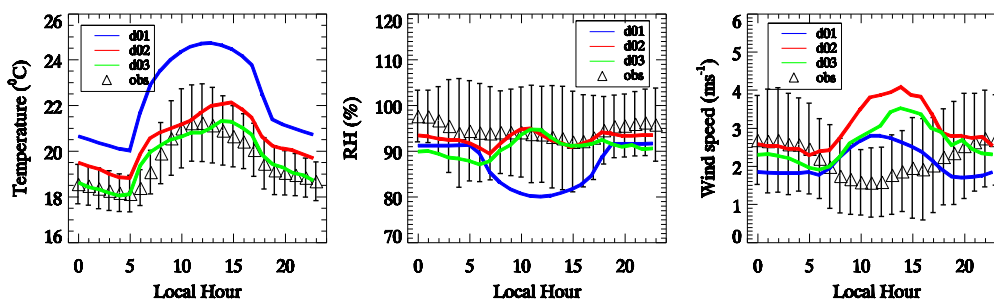
289 a correlation of $r = 0.75$ between d01 and observation. This warm bias is seen to decrease with increasing
 290 model resolution with bias reducing to 0.3°C only for the d03 simulation.

291 **Table1:** Mean Values of the meteorological parameters during summer monsoon (JJAS)

Parameter	Domain d01	Domain d02	Domain d03	Observations
Pressure (hPa)	869.5 ± 0.6	835.3 ± 0.6	827.6 ± 0.6	801.3 ± 0.3
Temperature ($^{\circ}\text{C}$)	22.3 ± 1.8	20.45 ± 1.8	19.8 ± 1.1	19.5 ± 1.1
Relative Humidity (%)	86.8 ± 4.9	92.7 ± 1.3	90.8 ± 2.0	94.5 ± 1.5
Wind Speed (ms^{-1})	2.1 ± 0.4	3.0 ± 0.6	2.6 ± 0.5	2.2 ± 0.4

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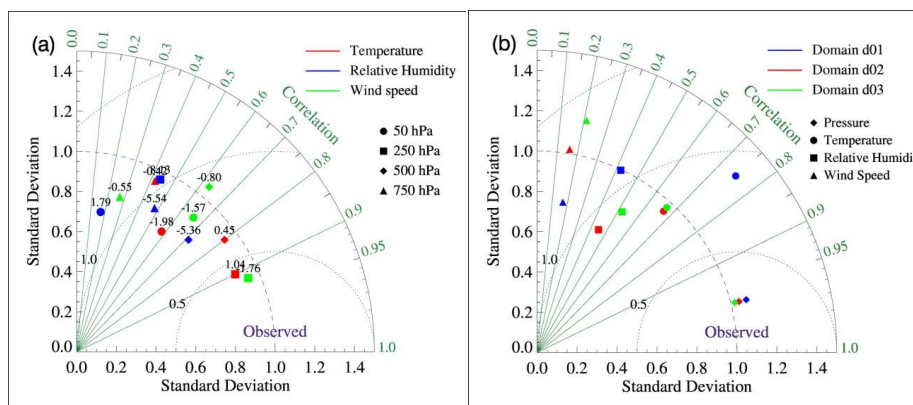


294

295 **Figure 5:** Mean diurnal variations in temperature, relative humidity, and wind speed from three
 296 simulations and observations.

297

298 The 10 m wind speed shows explicit trend of variation with grid resolution with finest resolution predicting
 299 nighttime wind speeds quite well. Diurnal mean wind speed is less predictable and shows an opposite trend
 300 of variation in daytime, which may be attributed to the orography induced effects. In case of the coarser
 301 resolution domain (d01), the wind speed is underestimated by 0.1 ms^{-1} while overestimated by 0.4 ms^{-1} in
 302 d03 with a poor correlation ($r \sim 0.21$).



303

304 **Figure 6:** Taylor diagram with correlation coefficient, normalized standard deviation, and normalized root
305 mean square difference (RMSD) error for (a) model performance at different pressure levels shown in
306 Figure 3 and (b) the model simulated surface pressure, 2m temperature, RH and 10 m wind speed for
307 different domains as shown in Figure 5.

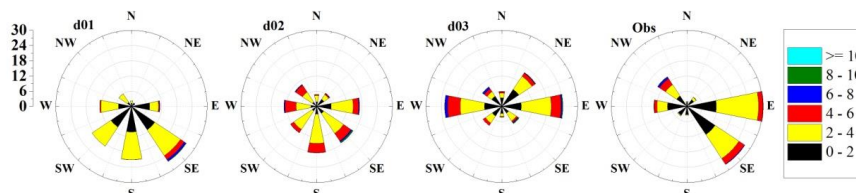
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309 The simulated relative humidity shows decreasing trend during daytime and reaches up to a minimum
310 value of about 79.9% between 11:00 to 13:00 LST for d01. Model simulates mean relative humidity of
311 $86.8 \pm 4.9\%$ with dry bias of 8% with moderate correlation ($r = 0.42$) for domain d01. The dryness in the
312 simulated atmosphere decreases up to 3% in d03 ($r = 0.52$) with the increasing grid resolution. Overall,
313 various aspects of the model performance such as normalized standard deviation, normalized root mean
314 square difference error and the correlation with ground-based observations are summarized in form of
315 Taylor's diagram in Figure 6b. From Table 1 and Figure 6b the model simulated surface pressure shows a
316 positive bias of 68 hPa with a good correlation of 0.97 in d01 with respect to the observed value which is
317 about 801.3 hPa. The bias decreases up to 26 hPa in d03 i.e. the highest resolution simulation.

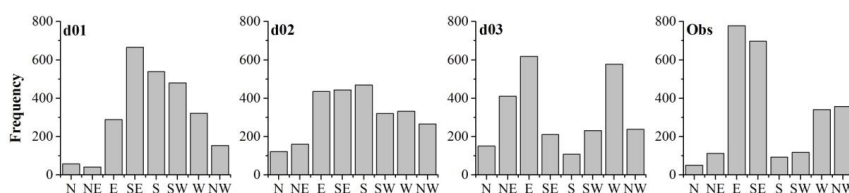
318 The normalized standard deviation of meteorological parameters as shown in Figure 6b explains that
319 except simulated 2 m temperature in d02 and wind speed in d03, all meteorological parameters are show



320 lower variability. As clear from Table.1, the temperature warm biases in d02 and d03 (0.9⁰C and 0.3⁰C
321 respectively) are lower as compared to the d01 (2.8⁰C).



322



323

324 **Figure 7:** Comparison of the wind direction in form of (a) wind-rose diagrams and (b) frequency
325 distribution from model simulations at three different resolution and observations

326

327 Wind directions are analysed in Figure 7 using wind-rose diagrams and frequency distribution. The winds
328 varying between meteorological direction 337.5^0 and 22.5^0 are considered to be the Northerly and
329 represented by N in the frequency distribution and so on for other directions taking into account the
330 clockwise meteorological convention. Wind direction are seen to vary differently at different model
331 resolutions. The frequency of southerly (539) and south-westerly (SW, 481) are quite higher in d01 as
332 compared to the observations (93 and 118 respectively), which decreases up to 109 and 232 in d03. Model
333 is able to simulate the northerly and north-easterly winds in d01 and d02, while, model simulates larger
334 contribution of north easterly winds in d03. The dominance of the summer monsoon seasonal easterly
335 (30%) and south-easterly (27%) winds is seen in the observation. The easterly component of the model
336 simulated wind shows better agreement with observations on increasing the model resolution.
337 Additionally, the model is able to simulate the westerly and north-westerly wind contribution in d02,

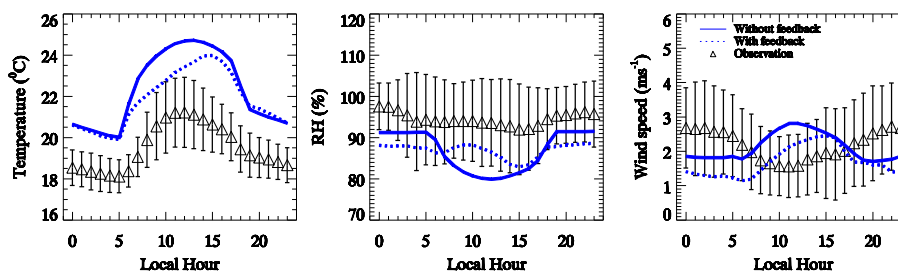


338 whereas, the westerly component is over predicted by 70% in d03. Hence, it is concluded that wind
339 direction also shows explicit dependency on the model resolution.

340

341 3.4. Effect of feedback

342 In the preceding section, the results of the simulations carried out without any feedbacks (WRF-WF) from
343 the finer resolution domain to its parent domain were presented. This WRF-WF experiment was conducted
344 in such a way that it could explicitly account for the grid resolution effects on the model performance. The
345 simulated meteorology with this model setup depicted different model performance in outermost coarse
346 resolution domain d01 as compared to d02 and d03 (Fig. 2 and 5). Another model simulation is carried
347 out in this section using the same configuration but with two-way interactive nesting and feedback (WRF-
348 F) from daughter domain to its mother domain. The model results over CH region in the regional scale
349 simulation (d01) shows better agreement with the observations because of the feedback from high
350 resolution nested simulation. The comparison of the simulated meteorological parameters (2m
351 temperature, RH, and 10 m wind speed) for outermost domain with the surface observations is presented
352 in Figure 8 for both WRF-WF and WRF-F, and the effect of the feedback within outermost domain is
353 given in Figure 9.



354

355 **Figure 8:** The diurnal variation of the 2 m temperature, relative humidity and 10 m wind speed.

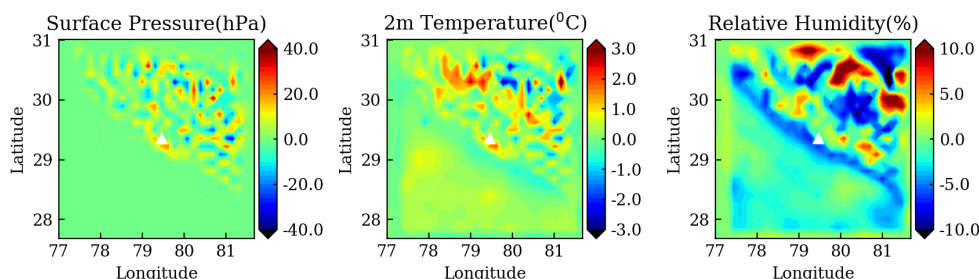
356



357 The analysis of simulated meteorological parameters for observational site in WRF-WF and WRF-F
358 experiments reveal the following:

- 359 1. The range of the diurnal variation of 2 m temperature in d01 changes from 20 – 24.7⁰C in WRF-
360 WF to 19.9-23.9⁰C in WRF-F with a slight decrease in warm bias from 2.8⁰C to 2.4⁰C (Figure 8).
361 This reduction in bias is the low temperature feature of the elevated terrain captured at higher
362 resolution which is feedback to the d01.
- 363 2. For observational site, WRF-F simulated relative humidity is 86.9% which is 0.1% wetter than
364 WRF-WF for outermost domain which is negligible. Also, WRF-F shows a different trend of
365 variation during daytime as compared to the WRF-WF for domain d01 (Figure 8). The comparison
366 of the WRF-WF and WRF-F simulated meteorology is compared in Table 2.
- 367 3. The effect of the feedback is more significant over the mountainous region than in the plains as
368 evident from the Figure 9.
- 369 4. Changes akin to 2 m temperature are observed in the surface pressure in WRF-F (Table 2). The
370 positive bias is reduced by 11 hPa. Again, the same low-pressure feature of elevated mountain
371 peaks is identified and captured by the model through the feedback.
- 372 5. Model simulated wind speed between the two cases: WRF-WF and WRF-F shows noticeable
373 differences, although the trend of diurnal variation remains similar. The WRF-F simulated wind
374 speed is lower by 0.4 ms⁻¹ in domain d01 as compared to WRF-WF, which is also close to the
375 observation (2.2±0.4ms⁻¹) made during the GVAX campaign.

376



377

378 **Figure 9:** The effect of the two-way nesting on d01 is shown. The difference between the simulations with
 379 feedback (WRF-F) and without feedback (WRF-WF) is shown.

380

381 **Table 2:** Comparison of the simulated meteorology in two model simulations: WRF-WF and WRF-F at
 382 the observational site from outermost domain.

Parameters	Observed value	WRF-WF		WRF-F	
		Value	Bias	Value	Bias
Pressure (hPa)	801.3±0.3	869.5±0.6	+68.2	858.9±0.7	+57.2
Temperature (°C)	19.5±1.1	22.3±1.8	+2.8	21.9±1.4	+2.4
RH (%)	94.5±1.5	86.8±4.9	-7.7	86.9±4.9	-7.6
WS (ms ⁻¹)	2.2±0.4	2.1±0.4	-0.1	1.7±0.4	-0.5

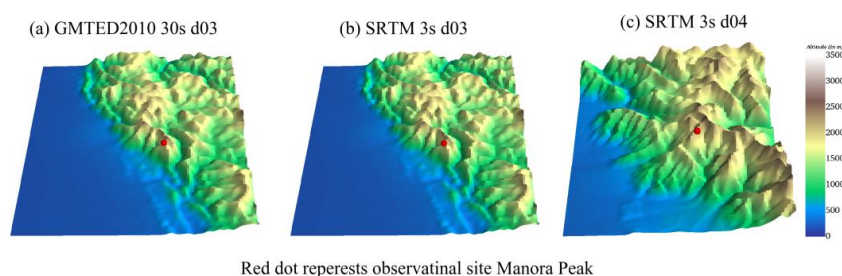
383

384 The feedback from the daughter domain to parent domain process mostly modifies the meteorology over
 385 the mountainous region within the domain. The effect of feedback is strikingly observed for the 2m
 386 temperature and the trend of diurnal variation of the relative humidity. Over all the model performance
 387 improved with the feedback, nevertheless further modelling studies alongside with more observations are
 388 needed to improve the model performance. Next, we implement a high resolution (3s) topographical input
 389 in the model to evaluate further fine resolution features over Himalaya.



390 **3.5. Inclusion of high resolution (3s) SRTM topography**

391 Simulations described in previous sections were performed using the 30s (~1km) topographic data from
392 the GMTED2010 (Danielson and Gesch, 2011). The resolution of this (30s or ~0.95km) is comparable to
393 the highest resolution of the WRF simulation (d03). To evaluate influences of topographical features over
394 even finer scales on the wind flows over this highly complex terrain, topography input available at very
395 high resolution (3s or ~90m) from the Shuttle Radar Topography Mission (SRTM3s) (Farr et al., 2007)
396 has been implemented without altering the model configuration, except performing the simulation as d04
397 (~333m). Simulation with SRTM data but at 1 km resolution did not differ significantly with the similar
398 resolution simulation using GMTED2010. For this experiment, model simulation is performed for 1 month
399 only (September 2011). This simulation carried out without feedback and compared with the observation
400 to check the effect of implementing high-resolution topography.



401

Red dot represents observational site Manora Peak

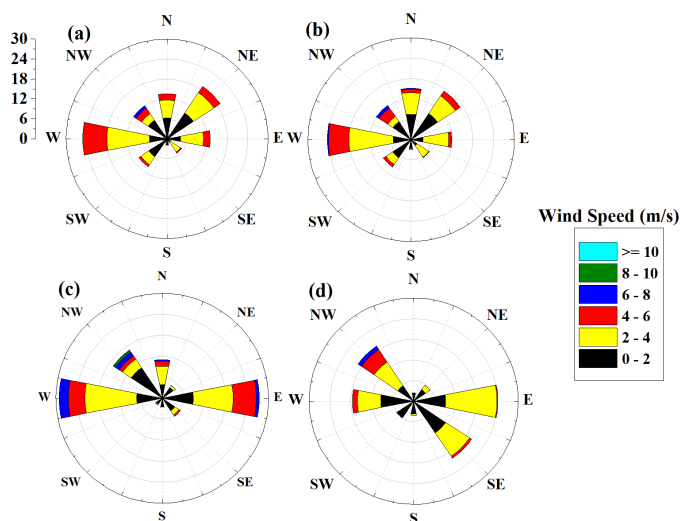
402 **Figure 10:** The topography from GMTED2010 at 30s and SRTM at 3s in domain d03 and d04.

403

404 The topography does not change much within d02 and d03 by changing input from GMTED2010 to
405 SRTM3s (Supplementary Figure S1). The topography in the d04 get better resolved as depicted by sharp
406 variations of mountain ridges and valleys using the SRTM3s as compared to the d03, as shown in the
407 Figure 10, which could be smoothed out if the simulation was carried out with GMTED2010 / or at 1 km
408 with SRTM3s. The induced effects due the more resolved topography in d04 can better simulate the local
409 circulation of air mass. Therefore, simulated 10 m wind direction in d04 is compared with the observations



410 and d03 to investigate the effect of including the SRTM3s topography. Surface pressure is seen to be
411 simulated more realistically (809 hPa) and the dry bias in 2m relative humidity is improved by ~2%.
412
413 Wind variations are shown in forms of the wind rose diagram using GMTED2010 and SRTM3s (Figure
414 11 and Supplementary Figure S2). The southerly wind component consistently shows an agreement with
415 the observations with increasing model resolution. The observation shows the prevalence of north-westerly
416 (19%), easterly (25%), westerly (18%) and south-easterly (20%) winds and these are also seen to be
417 dominant directions in the simulation d04, while occurrence of south-easterly winds is underestimated.
418 The least observed component is the northerly wind with (3%) while simulated by model is about 15%
419 and 11% in d03 and d04 respectively.



420
421 **Figure 11:** Wind rose diagrams for d03 using GMTED2010 (a), d03 and d04 using SRTM3s topography
422 data in (b), (c), and observation (d).
423
424 The observed southerly wind component is 4% while simulated as (3%) in d03 and d04. Simulation of the
425 wind directions improved from d03 to d04 by using the SRTM3s topography being relatively in better



426 agreement with observations, except certain wind directions such as south easterly. An improvement is
427 noticed in simulated surface pressure, 2 m relative humidity and 10 m wind speed using the SRTM3s
428 topography over the complex CH. The effects of the SRTM3s topographic static data is studied previously
429 over other regions of the world (e. g. Teixeira et al., 2014; De Meij and Vinuesa, 2014). The differences
430 between model and observations of winds over the Himalayan region are suggested to be associated with
431 still unresolved terrain features, besides the influences of input meteorological fields as well as the model
432 physics on simulated atmospheric flows (e. g. Xue et al., 2014; Vincent et al., 2015).

433

434 **4. Summary and Conclusions**

435 In this study, the effects of spatial resolution on model simulated meteorology over the CH has been
436 examined combining the WRF model with ground-based, balloon-borne observations during and intensive
437 field campaign, and reanalysis datasets. Owing to the highly complex topography of the central Himalaya,
438 model results show strong sensitivity towards the model resolution and adequate representation of terrain
439 features. Model simulated meteorological profiles do not show much dependency on resolution except in
440 the lower atmosphere, which is directly influenced by terrain induced effects and surface characteristics.
441 The biases in 2 m temperature, relative humidity and pressure show a decrease on increasing the model
442 resolution indicating a well resolved representation of topographical features. Diurnal variations in
443 meteorological parameters also show better agreements on increasing the grid resolution. Although the
444 surface pressure does not show a pronounced diurnal variation nevertheless the biases in simulated surface
445 pressure reduce drastically over fine resolution simulations. Model is generally not able to reproduce the
446 wind directions well, except some of the major components in all the simulations with varying resolutions.
447 A sensitivity experiment with domain feedback turned ON shows that the feedback process can improve
448 the representation of the CH in the simulation covering larger region of the northern Indian subcontinent.
449 It is suggested that further improvements in the model performance are limited due to the lack of high-
450 resolution topographical inputs, biases through input meteorological fields, and model physics.



451 Nevertheless, an implementation of a very high resolution (3s) topographical input using the SRTM data
452 shows potential to reduce the biases related to topographical features to some extent.

453

454 **Code and data availability**

455 WRF is an open-source and publicly available model, which can be downloaded at
456 http://www2.mmm.ucar.edu/wrf/users/download/get_source.html. Observations from the GVAX field
457 campaign are also available freely (<https://adc.arm.gov/discovery/#v/results/s/fsite::pgh.M>).

458

459 **Author contributions**

460 NS and AP designed and supervised the study. JS performed the simulations, assisted by NO and AS. JS,
461 NO, AS analysed the model results and NVPKK, KR, SSG contributed to the interpretations. VRK
462 contributed significantly in conceiving and realizing the GVAX campaign. JS and NS wrote the first draft,
463 and all the authors contributed to the manuscript.

464

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