Research Article

Validation of the HIRHAM-Simulated Indian Summer Monsoon Circulation

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The regional climate model HIRHAM has been applied over the Asian continent to simulate the Indian monsoon circulation under present-day conditions. The model is driven at the lateral and lower boundaries by European reanalysis (ERA40) data for the period from 1958 to 2001. Simulations with a horizontal resolution of 50 km are carried out to analyze the regional monsoon patterns. The focus in this paper is on the validation of the long-term summer monsoon climatology and its variability concerning circulation, temperature, and precipitation. Additionally, the monsoonal behavior in simulations for wet and dry years has been investigated and compared against several observational data sets. The results successfully reproduce the observations due to a realistic reproduction of topographic features. The simulated precipitation shows a better agreement with a high-resolution gridded precipitation data set over the central land areas of India and in the higher elevated Tibetan and Himalayan regions than ERA40.

1. Introduction

The Asian monsoon circulation is characterized by variability on seasonal, interannual, and interdecadal time scales and impacts on water availability in a highly populated region with dominating agriculture. For this reason, the understanding of the complex mechanisms and interactions between the atmospheric patterns affecting the entire monsoon system is of special interest. Global general circulation models (GCMs) are often used to simulate the large-scale circulation of the monsoon (e.g., [1–3]). Regional Climate Model (RCM) systems with higher spatial and temporal resolutions can add value at regional scales to the climate statistics when driven by GCMs with accurate large scales. Therefore, RCMs nested in a GCM or driven by data analyses have been applied for a special area of interest by dynamical downscaling and to improve their performance (e.g., [4, 5]). Several RCM simulations have been carried out for the South Asian Monsoon region (e.g., [6–13]).

In this study, we apply the regional climate model HIRHAM on an Asian integration domain and analyze a 44-year-long simulation 1958–2001 driven by ECMWF reanalysis (ERA40 data). The HIRHAM model is one state-of-the-art RCM and has been already successfully applied for European regions (e.g., [14, 15]), for the Arctic (e.g., [16–18]) as well as for Antarctica (e.g., [19, 20]). HIRHAM has been integrated over the Asian continent covering the high topography of Himalayas and Tibetan Plateau to simulate the Indian monsoon under present-day conditions. Before the model can be applied to improve the understanding of the regional coupling between monsoon circulation and orographic as well as thermal processes in the Himalayas and Tibetan Plateau, an essential prerequisite is the careful validation of the model simulations. Therefore, in the present study, the model integration over 44 years has been validated against observational data. The focus is on the validation of the summer monsoon circulation and the comparison of regional atmospheric circulation, temperature, and precipitation patterns between the HIRHAM model and the observational data. Related to this, the long-term climatology, the variability, and the monsoon behavior in selected wet and dry monsoon years have been evaluated. After a short introduction of the HIRHAM model and the used observational data sets (Section 2), the simulation results for mean sea level pressure, 2m air...
temperature and precipitation, geopotential and wind at different vertical levels, and a potential convective instability index are validated against the different data sets (Section 3). Finally, concluding remarks summarize the main results (Section 4).

2. Model and Observational Data

2.1. Regional Climate Model HIRHAM. The regional climate model HIRHAM used in this study has been developed by Christensen et al. [21]. It has been adapted to polar regions by Dethloff et al. [22] and Rinke et al. [23]. The dynamical part of the model is based on the hydrostatic limited-area model HIRLAM (High-Resolution Limited Area Model) [24] with prognostic equations for horizontal wind components, temperature, specific humidity, cloud water content, and surface pressure. The physical parameterizations for radiation, land surface processes, sea surface sea-ice processes, planetary boundary layer turbulence, gravity wave drag, cumulus convection, and large-scale condensation are taken from the general circulation model ECHAM4 [25]. Details of the model dynamics and physics can be found in the given references.

The integration domain in the Asian region (see Figure 1) covers an area between 0°N to 50°N and 42°E to 110°E, including the high topography of the Tibetan Plateau and the Himalayas as well as the northern Indian Ocean. The model domain includes 110 × 100 grid points and a horizontal resolution of 0.5°. The vertical discretization consists of 19 irregularly spaced levels in hybrid ζ-p coordinates from the surface up to 10 hPa. The model time step is 300 s.

The model has been run over the ERA40 period, that is, for 44 years from 1958 to 2001. At the lateral and lower boundaries, HIRHAM has been forced by ERA40 reanalysis data from the spatially interpolated 0.5° grid (see Section 2.2). The lateral forcing includes all prognostic variables except the cloud water content. The information from the lateral boundaries is transferred into the model interior by a boundary relaxation [26] in a 10-grid-point-wide boundary zone with boundary data updated 4 times a day. At the lower boundary, ECMWF-analyzed sea surface temperature has been used and updated daily.

2.2. Observational Data Sets. The model validation has been carried out by using gridded data from different sources, in which the data sets have been compared with the model output and with each other.

Table 1 contains all gridded data sets used for model validation. The data are characterized by a different spatial and temporal resolution. From the ERA40 reanalysis of the European Centre for Medium-Range Weather Forecasts (ECMWF) [27], the 2m air temperature, mean sea level pressure, precipitation, geopotential, and wind have been used. From the Global Precipitation Climatology Centre, the GPCC4 data set for precipitation has been applied; (ftp://ftp-anon.dwd.de/pub/data/gpcc/html/fulldata_download.html [28]). Additionally from the Indian Meteorological Department (IMD) [29], the precipitation data are analyzed.

2.3. Analysis Methods. In order to carry out direct comparisons between the gridded data sets and the model simulation results, the observations have been interpolated to the HIRHAM grid. For the quantitative comparison between HIRHAM and the gridded data, common statistical parameters (root mean square error (RMSE), bias “HIRHAM model minus observation” (BIAS), and pattern correlation coefficient (PATCOR)) have been calculated for the entire model area (ALL) as well as for subdomains with regionally specific criteria of selection (see Table 2: land surface (LAND), ocean (SEA), mountains with elevations higher than 1500 m a.s.l. (ELEV1500), and the Indian continent (INDIA)).

3. Results

The three-dimensional dynamical processes of the complex Indian monsoon circulation and its characteristic patterns (e.g., [30–32]) will be described in the following by an extensive validation of the model. The focus is on summer monsoon (June-September), 1958–2001.

3.1. Atmospheric Circulation. For the validation of the mean circulation patterns at different atmospheric levels, the seasonal mean ERA40 reanalysis data have been used. Figure 2 represents the mean sea level pressure and the 500-hPa geopotential height, averaged over the simulation period 1958–2001 for the summer monsoon season (JIAS).

The HIRHAM simulation shows an extended low pressure system in northwestern India, which covers the entire northern part of India, the Indo-Gangetic plains south of the Himalayas and the northern Indian Ocean with values below 1000 hPa. It is connected to the position of the near-surface monsoonal heat depression. Most parts of the domain are characterized by small differences “HIRHAM-ERA40”, ranging from −2 hPa to +2 hPa for mean sea
Figure 2: Mean sea level pressure (hPa; shaded) and geopotential height at 500 hPa (gpm; contour lines) for HIRHAM (a) and the difference “HIRHAM-ERA40” (b); summer monsoon (JJAS) 1958–2001.

Figure 3: Mean wind fields (wind vectors) and wind speed at 850 hPa (m/s; colours) for HIRHAM (a) and ERA40 (b); summer monsoon (JJAS) 1958–2001.

These differences in topography are reflected in larger circulation differences in mountain areas. The good agreement in the mean near-surface large-scale atmospheric circulation between HIRHAM and ERA40 within the entire integration domain is demonstrated by a high pattern correlation coefficient (0.95), a small bias (0.5 hPa), and a small RMSE (1.5 hPa) for the MSLP (see Table 2). The influence of the differently resolved orography between HIRHAM and ERA40 with resulting MSLP anomalies in the mountain regions becomes obvious by calculating the regional statistics for different subdomains. Compared to “ALL”, the BIAS (2.0 hPa) and RMSE (2.9 hPa) are higher for “ELEV1500”, while the pattern correlation is lower (0.80).
On the contrary, the BIAS (0.1 hPa) and RMSE (0.6 hPa) are smaller for “SEA” showing a higher PATCOR (0.99) than in the entire domain (see Table 2).

The lower troposphere is characterized by a strong low-level southwestern wind field onto the coastlines of South Asia across the Arabian Sea and the Bay of Bengal, representing the Cross Equatorial Jet (CEJ) in Figure 3. HIRHAM and ERA40 show very similar patterns in the climatological low-level (850 hPa) wind fields. The higher resolved HIRHAM model topography affects the regional wind systems in the highly elevated region.

The simulated geopotential heights at 300-hPa, shown in Figure 4, are affected by an upper tropospheric anticyclone located over the Tibetan Plateau. At its northern boundary, the Circumpolar Westerlies (CPW) are the significant flow patterns. South of that anticyclone, the upper tropospheric Tropical Easterly Jet (TEJ) occurs. All these patterns are well reproduced in HIRHAM, also visible in the comparison of the climatological upper troposphere 300 hPa wind fields shown in Figure 5. In comparison with ERA40 data, the geopotential differences in Figure 4 are small, within ±10 gpm for the geopotential height in 300 hPa and up to +20 gpm for the 70-hPa geopotential height in lower stratosphere.

Compared to ERA40 data, the mean atmospheric large-scale circulation is well reproduced by the HIRHAM model. Differences are mainly associated with a different horizontal resolution, where HIRHAM contributes with an added value.
3.2. Temperature. The near-surface 2m air temperature has been validated against the seasonal mean ERA40 reanalysis data averaged over the simulation period 1958–2001 for the summer monsoon season (JJAS). The highest summer temperatures, presented in Figure 6, occur in the Tharr desert in northwestern India and the Arabian Peninsula. A well-pronounced temperature gradient between the warm flat land areas of India and the colder elevated mountain regions of Tibetan Plateau is simulated. Small differences between −2 K and +2 K occur in most parts of the domain (Figure 6(b)). In the mountains, higher temperature differences (within ±8 K) are calculated, which is attributed to the differently resolved orography. The topography-related temperature differences between HIRHAM and ERA40 become also obvious by examining selected statistical parameters represented in Table 2. For the entire domain “ALL”, a high pattern correlation coefficient (0.96), a small bias (0.3 K) and a small RMSE (2.3 K) occur. In “ELEV1500”, larger values for BIAS (−1.7 K) and RMSE (3.5 K) and a lower PATCOR (0.90) appear, similar to that discussed for MSLP.

3.3. Precipitation

3.3.1. Climatology. Due to summer monsoon circulation and orographic uplifting processes, intensive and long-term rainfall events occur especially in the windward of the mountains and the offshore coastal regions. Therefore the maxima of accumulated summer precipitation with more than 800 mm are observed in the windward of the Western Ghats at the Malabar Coast, at the southern slopes of...
Himalayas (Khasi Mountains and Shillong Plateau) and in the southwestern windward of the Burman mountain ranges in the eastern Bay of Bengal. These three zones are well reproduced by HIRHAM (see Figure 7). On the other side, lee effects cause less rainfall amounts below 100 mm, seen for example in the Dekkan Plateau. The smallest amounts below 10 mm are simulated in the desert and semidesert areas. The precipitation validation has been carried out with seasonal mean ERA40 reanalysis data averaged over the simulation period 1958–2001 for the summer monsoon season (JJAS). Furthermore, GPCC4 and IMD-RF data have been used as additional gridded data sets.

In comparison with ERA40 (Figure 7(b)), mostly there are differences from −50 mm to +50 mm and higher differences in those regions, where the main part of summer precipitation occurs. HIRHAM underestimates precipitation at the southern slopes of Himalayas by ca. −200 mm and shows an overestimation by ca. +200 mm in the Khasi Mountains and Shillong Plateau southeastern of Himalayas near Assam, attributed to the smoothed orography in ERA40 due to the coarser horizontal resolution. The statistical parameters (see Table 2) show a relatively high pattern correlation coefficient (0.75), a BIAS of 27.4 mm, and a RMSE of 128.5 mm for “ALL”. The subdomains are characterized by a pronounced differentiation with higher BIAS and RMSE especially for “SEA”. There is a significant dipole pattern of dry precipitation differences located south of Ceylon and the southern Bay of Bengal with amounts below −200 mm and humid differences west of Malabar Coast over Arabian Sea and the central Bay of Bengal with more than +200 mm (see Figure 7(b)). That dipole pattern is associated with the weaker pattern correlation over “SEA” (see Table 2). To understand this anomaly pattern, a sensitivity study concerning the impact of the lateral boundary zone of different widths has been carried out (see the appendix).

Precipitation is strongly affected by topography [34], and that underlines the need for a high resolution in HIRHAM. Related to the better spatial resolution (0.5°) and the large number of assimilated land surface stations, GPCC4 is better suited to validate the HIRHAM-simulated precipitation than ERA40. There are smaller differences “HIRHAM-GPCC4” (Figure 7(c)) compared to “HIRHAM-ERA40”. The pronounced negative difference at the southern Himalayan slopes found in the comparison to ERA40 is almost missing. Therefore, the HIRHAM-simulated precipitation is closer to the GPCC4 precipitation in the higher elevated regions, verified by a lower bias (4.1 mm) and a lower RMSE (73.3 mm). Higher pattern correlations for “HIRHAM versus GPCC4” in “ELEV1500” (0.85) and “INDIA” (0.81) compared to “GPCC4 versus ERA40” in “ELEV1500” (0.60) and “HIRHAM versus ERA40” in “INDIA” (0.77) show superiority of high-resolution model that is the added value of the HIRHAM simulations (see Table 2).
Summer monsoon precipitation is partly affected by convective rainfall events. Therefore, dynamical processes, which impact on convection, have been analyzed. According to that an index of potential convective instability for the last decade (1990–2000) has been calculated for HIRHAM and ERA40, to describe and compare the dynamical mechanisms of the rainfall patterns in summer (see Figure 8). The index [35] is based on the difference of the equivalent potential temperature between 700 hPa and 1000 hPa. Positive values represent stable tropospheric conditions, and negative values indicate the potential convective instability of the lower troposphere, which is a crude indicator of convection and rainfall in summer. The entire Indian Ocean and the land surface south of Tibetan Plateau and Himalayas are characterized by instable conditions with minima in the northern Arabian Sea. Comparing the HIRHAM and ERA40 potential convective instability indices, the patterns in HIRHAM show more regional features over high topography due to the higher horizontal resolution. The central Indian land areas and the northern Arabian Sea are represented by a stronger instability than in ERA40 related to a higher potential convective activity characterized by positive anomalies in both “precipitation minus evaporation” and vertical velocity at 500 hPa and more total rainfall in this area. That result

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**Figure 9:** Mean annual cycles of precipitation (mm) 1958–2001 for HIRHAM (red), ERA40 (blue), and GPCC4 (green) in selected subdomains.
Table 2: Statistical parameters for HIRHAM model and observed data sets for mean sea level pressure (MSLP), 2m air temperature (T2m), and precipitation (Precip) in the entire model domain (ALL) and selected subdomains: mountain area >1500 m a.s.l. (ELEV1500), ocean (SEA), land surface (LAND) and Indian continent (INDIA), defined as pattern correlation coefficient (PATCOR), difference “HIRHAM model minus observation”, or the difference between two observational data sets (BIAS) and root mean square error (RMSE) in hPa (MSLP), K (T2m) and mm (Precip).

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The further inspection of the mean annual precipitation cycle shown in Figure 9 for selected subdomains (LAND, INDIA, TIBET, and GLACIER) over the long-term simulation period from 1958–2001 additionally emphasizes the added value of HIRHAM model against the driving ERA40 data. In “LAND” and “INDIA”, there is a good agreement between HIRHAM, ERA40, and GPCC4 with small differences throughout the year. In the summer months, HIRHAM simulates larger precipitation than ERA40 and is thus closer to GPCC4. In the higher elevated regions, “TIBET” and “GLACIER”, ERA40 shows a significant overestimation of precipitation particularly from May to September, that is, the whole summer monsoon season. Related to its better spatial resolution, HIRHAM is able to simulate the precipitation more realistically, compared to GPCC4. The statistical parameters computed in Table 2 underline that added value of HIRHAM.

Furthermore, the validation of HIRHAM-simulated rainfall over central India with another rain-gauge-based high-resolution gridded data set from the Indian Meteorological Department (IMD-RF) supports the already discussed added value of the HIRHAM simulation in that

agrees well with the increased rainfall over central India, which is closer to GPCC4 data set (see Figure 7) and demonstrates the added value of HIRHAM in those regions.
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Figure 11: Standard deviation of precipitation (mm; shaded) for HIRHAM (a), ERA40 (b), and GPCC4 (c); summer monsoon (JJAS) 1958–2001.

region (Figure 10). An area mean over a domain covering the characteristic rainfall pattern over central Indian land surface has been analyzed for IMD-RF, HIRHAM, and ERA40. HIRHAM is closer to the IMD-RF (266.1 mm) data by simulating more total precipitation (231.5 mm) than ERA40 (171.1 mm). The largest contribution to the rainfall originates from large-scale precipitation, and a smaller part is connected with convective precipitation. Both components are larger in the HIRHAM simulation compared to ERA40.

3.3.2. Interannual Variability and Decadal Means. The interannual variability, described by the standard deviation over the simulation period 1958–2001 for the summer monsoon season (JJAS) is presented for the precipitation in Figure 11. For a better comparison with GPCC4 land data set, only the values over the land area are shown. All data sets agree on a high year-to-year precipitation variability over the Western Ghats, central India, the southern slopes of Himalayas including Khasi Mountains and Shillong Plateau and over southeastern Asia. Compared to GPCC4, HIRHAM and ERA40 overestimate the interannual precipitation variability with values of up to more than 100 mm. A low variability is found over extratropical latitudes north of 30° N including Tibetan Plateau, over the Arabian Peninsula as well as over the desert and semidesert areas south of Caspian Sea.

Figure 12 represents the mean precipitation over the land areas for two selected 10-year periods within the whole 44-year-long HIRHAM simulation. The first decade (1970–1980) is according to [32] characterized by more wet and the second period (1990–2000) by normal summer monsoon rainfall conditions. In both decades, the HIRHAM precipitation resembles the GPCC4 data set quite well, especially over southern and central India, and over eastern Himalayas (Khasi Mountains and Shillong Plateau). Therefore, not only with respect to the long-term climatology but also on decadal time scales, the added value of HIRHAM compared to ERA40 has been demonstrated.

3.3.3. Wet and Dry Summer Monsoon Composites. After the successful validation of HIRHAM simulations with respect to the long-term and decadal climatology, extreme summer monsoon precipitation for individual years has been analyzed. The aim is to investigate the HIRHAM performance in simulating typical rainfall anomaly patterns between strong (wet) and weak (dry) monsoon (years). The selection of the anomalous wet and dry composites from the long-term simulation 1958–2001 is done by the criteria of Farrara and Yu [36]. Monsoon years are defined as anomalous wet (dry) if the summer (JJAS) precipitation averaged over the monsoon domain [35] is more (less) than 0.5 standard deviations above (below) the climatological mean.

In Figures 13(a)–13(c), the characteristic rainfall differences for “wet-dry” composite years (JJAS) are shown together with the same composites for ERA40 and GPCC4. The 95% statistical significance level of the differences is indicated by white contour lines. The HIRHAM simulations indicate a higher differentiation in the distribution of these patterns compared to ERA40. The negative precipitation anomaly over southern India is in better agreement comparing HIRHAM with GPCC4. With respect to the positive precipitation anomaly over central India and at the west coast of India, the HIRHAM simulation resembles the GPCC4 data set. The third positive precipitation anomaly found over south-eastern Asia is also reproduced by HIRHAM, while ERA40 fails to show it. On the other side, HIRHAM fails to reproduce the positive precipitation anomaly over the area of Bangladesh visible in GPCC4.

Figure 13(d) shows the “precipitation minus evaporation” and the vertical velocity at 500 hPa for “wet-dry” composite years (JJAS) based on HIRHAM simulations. The white dots represent the areas of strongest potential convective activity related to the occurrence of positive anomalies “precipitation minus evaporation” as well as positive anomalies of vertical velocity at 500 hPa, for “wet-dry”. It indicates that the three positive precipitation anomalies described in Figure 13(c) for the HIRHAM simulation are
associated with those regions indicating that the complex feedbacks between convective and large-scale precipitation are reproduced by the HIRHAM simulations to some extent. Moist convection is important over warm oceanic and continental regions that extend from the tropics into the mid-latitudes and includes individual clouds from horizontal scales between 1 and 10 km to mesoscale convective systems with horizontal scales around 1000 km. Convection affects the surface locally through heavy rains and through its aggregated effects on the large-scale atmospheric fluxes of momentum, moisture, and heat. Many atmospheric processes of varying scale influence the vertical thermodynamical atmospheric structure, in which convection is initiated. We computed a potential convective instability index, which is sensitive to the lower tropospheric moisture. The moisture content evolves according to a mixing ratio equation for

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**Figure 12:** Precipitation (mm; shaded) for HIRHAM (left), ERA40 (middle) and GPCC4 (right); summer monsoon (JJAS) 1970–1980 (a), 1990–2000 (b) and the difference “1970–1980 minus 1990–2000” (c).
Figure 13: Precipitation “wet-dry” (mm; shaded) and statistical significance (95%; white contour lines) for GPCC4 (a), ERA40 (b), and HIRHAM (c). Precipitation-Evaporation (P-E) “wet-dry” (mm; shaded) and vertical velocity (ω) at 500 hPa “wet-dry” (hPa/h; black contour lines) for HIRHAM (d); summer monsoon (JJAS) 1958–2001. White dots in (d) show areas with positive anomalies in both P-E and ω.

4. Conclusions

Simulations of the Indian summer monsoon circulation with the regional climate model HIRHAM have been represented and validated against several observational gridded data sets. The performance of HIRHAM concerning the climatology and variability of atmospheric circulation, temperature, and precipitation has been quantified. A rather good agreement between model and ERA40 data in the circulation and temperature patterns has been found.

The validation of rainfall climatology and variability with several observational data sets (ERA40, GPCC4 and IMD-RF) indicated that the simulations show a better agreement with the higher-resolution data sets and can deliver an added value compared to ERA40. The improvement is due to spatial resolution and the model physics. HIRHAM delivers the best agreement with the GPCC4 precipitation data set. The HIRHAM-simulated precipitation is closer to the GPCC4 precipitation in the higher elevated regions. For example, the pronounced negative difference at the southern Himalayan slopes, which was obvious in comparison to

water vapour. The local change of the mean mixing ratio depends on the horizontal and vertical advection of its mean components and on the horizontal and vertical flux convergences of its eddy component. Persistent large-scale ascent can saturate lower tropospheric layers. In this way, vertical motions induced by advective processes contribute to thermodynamical destabilisation in the vertical column. In the long-term mean, P-E is a measure of the column convergence of moisture, which is influenced by convection and the large-scale horizontal advection of moisture.
ERA40, is almost missing in the GPCC4 comparison. Over central India, there is a humid anomaly pattern “HIRHAM-ERA40”, which becomes smaller in “HIRHAM-GPCC4”. A higher pattern correlation for “HIRHAM versus GPCC4” in “INDIA” compared to “HIRHAM versus ERA40” shows the added value of the HIRHAM simulations over central India. The validation of the HIRHAM-simulated rainfall over central India with the rain-gauge-based high-resolution data set from the Indian Meteorological Department (IMD-RF) supports the discussed benefit of the HIRHAM simulation in that region.

Comparing the HIRHAM and ERA40 potential convective instability indices, the patterns in HIRHAM represent more regional features over the high topography due to
the higher horizontal resolution. The central Indian land areas and the northern Arabian Sea are represented by a stronger instability than in ERA40 related to stronger potential convective activity characterized by positive anomalies “precipitation minus evaporation” as well as positive anomalies of vertical velocity at 500 hPa and more total rainfall in this area. That result agrees well with the increased rainfall over central India, which is closer to GPCC4 data set.

The mean annual precipitation cycle has been investigated for selected subdomains over the long-term simulation period. In the higher elevated regions, “TIBET” and “GLACIER”, ERA40 shows a significant overestimation of precipitation particularly from May to September. Related to its better spatial resolution, HIRHAM is able to simulate this precipitation more realistically, compared to GPCC4.

All data sets agree on a high year-to-year precipitation variability over the Western Ghats, central India, the southern slopes of Himalayas, and over southeastern Asia. Compared to GPCC4, HIRHAM and ERA40 overestimate the interannual precipitation variability. A low variability is found over extratropical latitudes north of 30°N, well represented in all data.

After the successful validation of HIRHAM simulations with respect to the long-term and decadal climatology, the typical rainfall anomaly patterns between strong (wet) and weak (dry) monsoons (years) have been evaluated. The HIRHAM simulations indicate a higher differentiation in the distribution of these patterns compared to ERA40. The negative precipitation anomaly over southern India is in better agreement comparing HIRHAM with GPCC4. With respect to the positive precipitation anomaly over central India and at the west coast of India, the HIRHAM simulation resembles the GPCC4 data set. The third positive precipitation anomaly found over south-eastern Asia is also reproduced by HIRHAM, while ERA40 fails to show it. On the other side, HIRHAM fails to reproduce the positive precipitation anomaly over the area of Bangladesh visible in GPCC4. The simulated three positive precipitation anomalies are associated with areas of strongest potential convective activity characterized by positive anomalies in both P-E and vertical velocity at 500 hPa indicating that the complex feedbacks between convective and large-scale precipitation are reproduced by the HIRHAM simulations to some extent.

Appendix

Although we did not change the size of the integration domain like Castro et al. [4], this sensitivity study helps to understand the impact of different wide boundary zones on the simulation results. In an experiment (called HIRHAM “BOUND20”), the boundary relaxation has been expanded to a 20-grid-point-wide zone. The aim of this sensitivity study was to test its impact on the simulation results, since a significant dipole bias in the HIRHAM-simulated rainfall over the Indian Ocean near the equator occurred. A second aim was to study the impact of the changed boundary zone on the simulated convection and related rainfall patterns as well as the behaviour of wind fields near the southern boundary.

Figure 14(a) shows a significant weakening and northern shift of the rainfall dipole pattern. This demonstrates that part of the precipitation bias is connected with the lateral boundary forcing and the width of the boundary zone. According to other RCM studies (e.g., [6, 8–11, 37]), it is possible that another part contributing to the rainfall bias is associated with the cumulus convection parameterization. To explore this in detail is beyond the scope of this paper, but such HIRHAM sensitivity studies are planned for the future.

The characteristic rainfall pattern over central India (Figure 14(a)) and the distribution of strong convection over land surface (Figure 14(b)) are still present and independent of the applied boundary zone width.

Finally, the effect of the modified boundary zone on the wind fields at 850 hPa shows a northern shift of the maximum monsoonal wind flow across the central and southern Arabian Sea, India, and the Bay of Bengal (see Figure 14(c)) with impacts on the northern shift of the precipitation anomaly dipole pattern over sea (see Figure 14(a)). The characteristic high gradient of wind speed, which normally occurs over southern India and the Bay of Bengal (see “BOUND10”), is also shifted to the north, where it is weakening concurrently.

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