On the structure and dynamics of Indian monsoon depressions


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On the Structure and Dynamics of Indian Monsoon Depressions

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ABSTRACT

ERA-Interim reanalysis data from the past 35 years have been used with a newly developed feature tracking algorithm to identify Indian monsoon depressions originating in or near the Bay of Bengal. These were then rotated, centralized, and combined to give a fully three-dimensional 106-depression composite structure—a considerably larger sample than any previous detailed study on monsoon depressions and their structure. Many known features of depression structure are confirmed, particularly the existence of a maximum to the southwest of the center in rainfall and other fields and a westward axial tilt in others. Additionally, the depressions are found to have significant asymmetry owing to the presence of the Himalayas, a bimodal midtropospheric potential vorticity core, a separation into thermally cold (\(-1.5\) K) and neutral (\(-0\) K) cores near the surface with distinct properties, and the center has very large CAPE and very small CIN. Variability as a function of background state has also been explored, with land–coast–sea, diurnal, ENSO, active–break, and Indian Ocean dipole contrasts considered. Depressions are found to be markedly stronger during the active phase of the monsoon, as well as during La Niña. Depressions on land are shown to be more intense and more tightly constrained to the central axis. A detailed schematic diagram of a vertical cross section through a composite depression is also presented, showing its inherent asymmetric structure.

1. Introduction

The Indian monsoon trough region experiences 3–6 synoptic-scale cyclonic depressions passing through it during the average summer. The majority of these depressions originate over the head of the Bay of Bengal, usually propagating northward onto the Indian subcontinent, lasting an average of 4–5 days (Godbole 1977; Saha et al. 1981; Sarker and Choudhary 1988; Stano et al. 2002).

The India Meteorological Department imposes the criterion that depressions must have surface winds greater than 8.5 m s\(^{-1}\), or they are considered a low, and less than 16.5 m s\(^{-1}\), else they are considered a cyclonic storm (Saha et al. 1981; Krishnamurthy and Ajayamohan 2010); our discussion shall be limited to depressions, not including lows or cyclonic storms. They are responsible for heavy and extreme precipitation events as well as strong winds.

Detailed investigation of depression structure in and around the Indian monsoon trough region has been performed previously (e.g., Godbole 1977; Ding 1981; Ding et al. 1984; Sarker and Choudhary 1988; Prasad et al. 1990; Stano et al. 2002). Although the largest
composite generated was of 40 depressions (split into four categories of intensity) by Prasad et al. (1990), they only considered wind and included analysis of satellite-imaged cloud cover on a case-study basis. Sarker and Choudhary (1988) analyzed a number of variables (temperature, moisture, winds, vorticity) of a 27-depression composite based on events during 1961–74, but their data were interpolated from a relatively sparse array of radiosonde stations, all over land. More recently, Stano et al. (2002) conducted hydrometeor analysis on three depressions from 1999.

With the advent of extensive satellite and reanalysis datasets, a truly thorough analysis of a large depression composite over land and sea has now become a possibility. Hurley and Boos (2015) considered a 117-depression composite for 1979–2013, but they did not go into specific detail, instead considering the worldwide climatology of monsoon depressions and examining the winds, potential temperature, and potential vorticity of Indian monsoon depressions. Boos et al. (2015) looked at a potential vorticity analysis of the same dataset. Their analysis also contains more depressions because they consider those originating over the Arabian Sea and inland, whereas we do not.

Here, newly developed feature tracking software was applied to reanalysis data over 36 years (1979–2014), recovering 106 depressions (Fig. 1a) that generally initiated in the Bay of Bengal and made landfall on the Indian subcontinent. These tracks were corroborated with the Indian Meteorology Department (IMD) Cyclone and Depression eAtlas (http://www.imdchennai.gov.in/cyclone_eatlas.htm). Thirty-four of these occurred between 1999 and 2013 and thus fall within the Tropical Rainfall Measuring Mission (TRMM) regime (Kummerow et al. 1998).

We will analyze horizontal and vertical structure of the composite depression and compare the structure and dynamics of depressions when situated over ocean and land, as well as the influence of large-scale forcing from modes of intraseasonal and interannual variability on depression strength and structure. An outline of the data and methodology is given in section 2. We analyze horizontal cross sections of the composite monsoon depression in section 3, looking at both satellite-based and reanalysis data. In section 4 we then look at vertical cross sections of the composite.

Subsequently, we look at local variability within the composite in section 5 and also look at how intraseasonal and interannual modes of variability affect depressions in section 6, investigating the effects of ENSO (section 6a), the Indian Ocean dipole (section 6b), and the active–break cycle of the monsoon itself (section 6c). We finally conclude in section 7.

2. Methodology

a. Data

1) TROPICAL RAINFALL MEASURING MISSION

The Tropical Rainfall Measuring Mission comprises four satellite-based precipitation-related instruments that have been in orbit since November 1997, from which both raw and gridded datasets are produced (Kozu et al. 2001). For our analysis, we use the 3B42 gridded product (version 7), which derives from the onboard IR scanner. The product consists of 3-hourly gridded 0.25° × 0.25° surface precipitation over the latitude range 50°N–50°S (Huffman et al. 2007, 2010; Liu...
et al. 2012), with a nominal intensity resolution of 0.7 mm h^{-1} (Kozu et al. 2001). The advantage of the 3B42 algorithm product is its calibration and merging with products from a number of other IR-based precipitation satellites: Geostationary Meteorological Satellite (GMS), GOES-East, GOES-West, Meteosat-7, Meteosat-5, and NOAA-12. When compared to a climatology of surface observations, TRMM 3B42 performs well, although it tends to underestimate both low- and high-rainfall rates, generally slightly overestimating overall (Prakash and Gairola 2014) and significantly underestimate rainfall over the Western Ghats (Nair et al. 2009).

2) ERA-INTERIM REANALYSIS

European Centre for Medium-Range Weather Forecasts ERA-Interim reanalysis data (Dee et al. 2011) that we use are the surface analysis for horizontal compositing and the pressure level analysis for both horizontal and vertical (see section 4) compositing. Our analysis uses the 27 output pressure levels from 1000 to 100 hPa and the horizontal output resolution of approximately 0.7° × 0.7° (N128). In this reanalysis, the cloud cover field is purely model derived—it is calculated by summing the contributions of stratiform and convective cloud covers, respectively. Stratiform cover is predicted from local relative humidity (Sundqvist et al. 1989), whereas convective cover is predicted from parameterization of convection (Slingo 1987).

b. Tracking software

To automate the collection of data on depressions, feature tracking software was written, as summarized in Fig. 2. Because of the complex orography north of the trough and the sea-to-land transitions of depressions, we could not rely solely on any particular criterion, so a filtering procedure was carried out on relative vorticity, geopotential height, and wind speed.

For each 6-hourly time step, the tracking software finds a relative vorticity maximum at 800 hPa, above a threshold of 1 × 10^{-5} s^{-1}. It then eliminates all cases without a nearby synoptic-scale surface low (negative surface pressure anomaly relative to a 21-day climatology), or failing the wind threshold criterion for depressions. Use of the 800-hPa level reduces boundary layer effects from the Himalayas while sampling the vertical core. Consecutive reanalysis outputs containing depression candidates are then connected, as long as the candidate has not moved too far between frames, and locations, times, and headings are recorded.

Filtering is applied to ensure consecutive relative vorticity (ζ) maxima are attributable to the same event and that the events last a minimum of 24h. Before further analysis, the 6-hourly fields are smoothed with a 12-h moving window to reduce noise. Those depressions that were not in the IMD eAtlas were then removed. For verification, monsoon depressions from the Hurley and Boos (2015) dataset with the same spatial criteria are shown in Fig. 1b, with whom, for the same period, we share some 90% of events. The tracks from their dataset appear longer because they also include the parts of depression tracks where they were only strong enough to be considered a monsoon low (as demonstrated by the coloring). Prajeesh et al. (2013) considered both the IMD depression archive and the dataset of Sikka (2006) and showed that in each, a statistically significant decline in frequency exists. This was challenged, however, by Cohen and Boos (2014), who used several independent datasets (including that of Hurley and Boos 2015) to show that this might not be case, consequently calling into question the validity of the downward trend in the IMD dataset.

The mean depression heading in our composite was 332°; neglecting eastward-propagating depressions gives a mean heading of 295°. The mean propagation speed was 2.75 m s^{-1}, and all depressions that fit our criteria initiate and terminate between 11° and 27°N, and between 62° and 94°E.
Next, each depression was centralized to $0^\circ$ latitude, $0^\circ$ longitude and reoriented using output heading data to create a forward (northward) propagating composite. Rotation during compositing allows us to determine system-relative features and mitigates orographic artifacts (e.g., forced ascent by the Eastern Ghats). We use the term “relative” with compass directions to describe sectors of the depression; relative north being the direction of propagation. The resulting array was then sliced horizontally through different pressure levels to reveal internal structure of the composite depression. Apart from variables that are usually defined at the surface, the chosen pressure levels are usually 850 and 200 hPa.

c. Significance testing

We have sufficient data to perform significance testing on the total composite and an intercomparison within the composite to ensure the robustness of our results.

1) A Student’s $t$ test is performed to see whether the composite state significantly differs from the mean state (i.e., the summer climatology). Areas where a 95% confidence threshold is not satisfied are colored gray in these figures.

2) Subsets are tested against their superset composite (e.g., if we propose that a variable $\chi$ can be meaningfully split into subset fields $\chi_i$, we would test $\chi_i$ against $\chi$, rather than the background climatology). This testing is performed by bootstrap resampling the superset (see, e.g., Efron 1979) and producing 10,000 random samples replicating the size of the subset. The subset and random samples are then individually composited and each grid point in the subset is placed in the respective gridpoint distribution from the random samples. For 95% confidence that the subset is significantly different from its superset, we must return a percentile either below 2.5% or above 97.5%.

3) On pairs of subsets we perform an analogous method, except the bootstrapping produces 10,000 pairs of random samples (again, each with the same sample size) and differences each pair to produce the distribution of differences for comparison with the observed intersubset differences.

d. Choosing a suitable climatology

It is important to choose an appropriate climatology (e.g., a summer mean, a daily mean, or a fixed-length running mean) for our analysis on depressions as anomalies to the monsoon background state. When considering the full composite, as in sections 3 and 4, we bear in mind that not having a significant trend in depression frequency across the summer entitles us to make an arbitrary choice of climatology. This was confirmed with several sensitivity experiments on the composite (not shown). When we discuss time-dependent large-scale forcings, this assertion is no longer valid; thus, if we are to use a climatology in that analysis, it must be a running mean one so as to capture the background effect of the forcing.

3. Composite horizontal structure

In this section we describe the horizontal composite structure of the depressions, first looking at precipitation and then reanalysis fields.

a. Precipitation

TRMM precipitation data were composited for 1998–2013, the range of the dataset; Fig. 3a shows the average
precipitation for all 34 depressions that initiated in the Bay of Bengal in this period. The most intense rainfall rate found during any 3-hourly TRMM time step was 68.4 mm h\(^{-1}\) (1.6 m day\(^{-1}\) equivalent). Comparing this to the highest “daily mean” intensity in the 34-depression composite (38.6 mm day\(^{-1}\)) shows the sparsity and unpredictability of such intense events.

The location of the average rainfall maximum to the relative southwest agrees with Godbole (1977) and Yoon and Chen (2005). That the rainfall is most intense to the left of the central track has been known for some time (Ramanathan and Ramakrishnan 1933). Mooley (1973) provided a good summary of the literature on why this might be the case, including hypotheses from Roy and Roy (1930), Ramanathan and Ramakrishnan (1933), Mull and Rao (1949), Desai (1951), and Petterssen (1956). He came to the conclusion that a combination of factors are responsible: (i) the southwest axial tilt of the depression core with height, (ii) a convergence maximum in the relative west-southwest sector, and (iii) cyclonic mixing of warm, southwesterly air from the Arabian Sea with the generally cooler, easterly monsoon flow. Inspection of the distribution for “extreme” rainfall events (not shown) indicates that the probability for higher rainfall rates is increased near the composite maximum, although the highest intensities are seen sporadically outside this area and are probably usually unrelated to the depression itself.

Figure 3b shows the total rainfall across India for depression days, and Fig. 3c shows the anomaly of this relative to the summer mean for the TRMM period. The wettest anomalies are found around coastal areas (particularly the Bay of Bengal), and wet anomalies persist across much of the trough region, displaced from the track density (see Fig. 1a) several degrees to the southwest, as the precipitation structure (see Fig. 3a) predicts. There are also dry anomalies over the Himalayan foothills and northern Indian Ocean, creating something of a “wet” channel with dry edges, implying that a passing depression drags in surrounding moist air from a large area where precipitation is suppressed by atmospheric subsidence.

b. Reanalysis fields

We next explore dynamic and thermodynamic fields from reanalysis to help elucidate the depression structure.

First we will examine wind: Fig. 4 shows 850-hPa wind speed and direction for the composite depression (Fig. 4a). We see the maximum wind intensity is in the relative southeast, approximately 3° away from the center, whereas a local minimum occurs at the center (along with the streamfunction minimum; not shown), as in the eyes of more intense systems such as tropical cyclones. Removal of the background climatological winds (Fig. 4b) shows the perturbation the depression makes to the monsoon: the structure is more circular since the asymmetrical presence of the climatological monsoon winds at 850 hPa is now absent—particularly in the relative south and west sectors, and we see an intensification of the winds in the relative east due to the presence of the Himalayas via the image vortex mechanism described in
Hunt and Parker (2015, manuscript submitted to *Quart. J. Roy. Meteor. Soc.*). The overall wind speed pattern is very similar in shape, although with reduced intensity, to that described in Catto et al. (2010) for extratropical cyclones and is also similar to that of the smaller composites described in Godbole (1977) and Prasad et al. (1990).

Above 200 hPa (not shown), the depression contributes a weak (<4 m s\(^{-1}\)) anticyclone to the much stronger (>20 m s\(^{-1}\)) monsoonal jets at this height. There is an upper-level wind speed minimum above the core, as there is above tropical cyclones (DeMaria 1996). The background flow provides the vertical shear that restricts further development of the depressions into cyclones in midsummer (e.g., Ramage 1959; DeMaria 1996) so that tropical cyclones in the northern Indian Ocean have a bimodal distribution with maxima in May and November and minima in February and August (Kikuchi and Wang 2010).

We can now examine two more of the variables used in tracking the depressions: relative vorticity and pressure (Fig. 5). At 850 hPa (Fig. 5a), the negative relative vorticity imparted to the depressions by the presence of the Himalayas (friction and vortex squashing) is notable (pink contour level) in the relative vorticity field. Indeed, depressions at latitudes nearing the foothills are shorter lived (Fig. 1a), and north of a certain latitude, only depressions from the more favorable La Niña years (not shown) tend to persist, since the Bay of Bengal has greater relative vorticity and relative humidity (Felton et al. 2013). The relation between depression variability and ENSO will be discussed further in section 6a.

At 200 hPa (Fig. 5b) the expected anticyclonic pattern is clear: only the immediate center has positive relative vorticity, and from just a few degrees out this becomes negative (anticyclonic), again with influence of the Himalayas in the relative northeast sector where vorticity is most negative. This is further supported by the presence of a large shallow area of high geopotential over the relative northeast sector.

Two horizontal temperature slices are given in Fig. 6. At the surface, the low-level cold-core nature of the composite is clear, with a temperature minimum in the relative southwest sector. There is a consistently steep rise in temperature away from the center until a radius of approximately 7\(^\circ\), eventually becoming a slightly anomalously warm environment. At 350 hPa, the anomaly has opposite sign, becoming warm compared to the climatology. This warm maximum is located slightly relative southward of the center and has similar magnitude and gradient to its surface companion. However, there is greater asymmetry: the anomaly is nearly half a degree colder in the lee of the depression than ahead of it.

We now turn to relative humidity and analyzed cloud cover (Fig. 7). This total cloud cover bears some similarity with that of the extratropical cyclones in the atlas defined by Dacre et al. (2012). There is a maximum composite cloud cover of 87% in the relative southwest sector, near the precipitation maximum. There is greater cover in the lee of the depression, likely owing to the larger quantity of moisture and stronger surface winds there than in the advance of the depression.
The 850-hPa relative humidity field features a maximum in the relative southwest sector, as with rainfall and cloud cover. We see that the humidity falls away more rapidly ahead of the depression than behind it, since the depressions tend to have the Bay of Bengal environment behind them and the drier Indian subcontinent ahead.

We also analyzed surface heat exchanges and boundary layer processes (not shown); we caution that these are entirely modeled products within the reanalysis and thus have higher uncertainty than previous composite fields. However, our findings are in line with theoretical results: we find the boundary layer height drops sharply near the center to around 500 m, from the climatological value of 800 m. This finding couples with a reduction in magnitude of both sensible and latent surface heat fluxes: values are reduced from the typical climatologies of \(30\) and \(100\) W m\(^{-2}\) to \(8\) and \(75\) W m\(^{-2}\) over the center, respectively.

4. Composite vertical structure

Stano et al. (2002) estimate the vertical extent of a monsoon depression to be in the region of 12–15 km (200–120-hPa pressure height) based on radar hydrometeor data from TRMM, so we shall look at vertical cross sections over the 27 available ERA-Interim pressure levels from 1000 to 100 hPa. We start by considering the overall 106-depression composite before looking at the effects of larger-scale monsoon variability. ERA-Interim repeats values for the lowest atmospheric level over orography at all subsurface levels down to 1000 hPa; those data are not composited. The vertical composites are drawn above the zero-latitude line in a horizontal composite using the central grid point in the north–south direction, with relative west on the left and relative east on the right.

a. Moisture

Analysis of the vertical distribution and transport of moisture in the composite serves to verify the rainfall structure and provide further insight into the moist mechanisms of the depression.

Relative humidity (Fig. 8a) slightly resembles the structure of cloud cover: two strong maxima, one just above the surface at the center, and one much nearer the tropopause. There is again (as with, e.g., divergence) a relative westward axial tilt persisting until the axial minimum at approximately 350 hPa. The 90% isohume near the surface is large, spanning a cross-sectional area of approximately 1500 km\(^2\) (a volume of over 800 000 km\(^3\) assuming azimuthal symmetry), indicating the composite depression has a vast, nearly saturated core within it.

It is well known (e.g., LeMone et al. 1998; Cetrone and Houze 2006) that the tropical atmosphere has a high relative humidity, so we should explore exactly how much moisture the depression adds to the environment. The composite structure of relative humidity as an anomaly to the climatology (not shown) again shows westward tilt with height, corroborating the rainfall asymmetry discussed in section 3a. In the middle and lower troposphere,
the depression contributes around 10%–15%, resulting in a nearly saturated environment.

b. Cloud features

Figure 8b shows the vertical structure of cloud water content (CWC) across the depression. Cloud products from ERA-Interim are heavily reliant on the reanalysis model, but since they are constrained by other observations we may have some confidence in their patterns. CWC is a proxy for cloud type (Hess et al. 1998; Rosenfeld and Lensky 1998), with the threshold for cumulonimbus at $7 \times 10^{-5} \text{kg kg}^{-1}$ and stratocumulus at $3.5 \times 10^{-5} \text{kg kg}^{-1}$. We can identify a cumulonimbus base at about 900 hPa, and at around 550 hPa clouds are clearly more dense away from the depression center, again lending evidence to a weak, incomplete eyewall structure. Above approximately 450 hPa, most cloud mass comprises ice. The vertical structure of cloud cover (Fig. 8c) reveals two distinct regions of large cloud cover: a maximum near 850 hPa surrounding the center, and an upper-level maximum, to the relative west of the center, near 150 hPa. These are the clouds that dominate the visible-light satellite pictures, anticyclonically rotating cirrus, and cirrostratus clouds (Houze 2010). The general shape resembles that of a classic cumulonimbus cross section (with resolved low-level cloud and an anvil fed by deep convection/ascent), and there are potential rainbands visible in the relative east flank.

c. Potential vorticity and circulation

The vertical structure of potential vorticity (Fig. 9a) has a bimodal maximum core of a little over $10^{-6} \text{K m}^2\text{kg}^{-1}\text{s}^{-1}$ extending from approximately 750 to 450 hPa. There is a minimum aloft, at approximately 250–200 hPa before the effect of the high-PV stratosphere becomes apparent.

Tangential wind speed anomaly (Fig. 9b) follows approximately what we might expect from a classical cyclonic structure, with two lobes of local maximum intensity found slightly above the surface (~850 hPa) and at a distance of several hundred kilometers from a relatively calm center. The axial tilt to the relative west is mentioned both above and in Godbole (1977). Approaching the tropopause, the anticyclone is considerably affected by the strong zonal monsoon flow (~20 m s$^{-1}$), and the axis starts to tilt instead to the relative northeast (corroborated by the horizontal slice at 200 hPa; see
When the climatology is not removed, the tangential wind speed structure strongly resembles that shown in Sikka (1977) based on work by Sikka and Paul (1975) on a composite of July/August depressions in 1966–70 and the composite of Godbole (1977).

The vertical structure of relative vorticity anomaly is given in Fig. 9c. It has a very symmetric form, with a maximum intensity found at approximately 900 hPa. Godbole (1977) reported a similar shape, but with a slightly more elevated maximum, at approximately 800 hPa. Also visible is the very slight relative westward axial tilt also observed by Godbole (1977). The anticyclonic behavior expected near the tropopause is apparent away from the center (as seen in Fig. 5b). Again, the imposition of negative vorticity from the Himalayas in the relative east is visible.

Vertical velocity (Fig. 10) has an axial tilt and slightly bilobal structure extending into the midtroposphere. There is weak ascent at the surface, but parcels are rapidly accelerated near the core, reaching a peak of over 0.2 Pa s\(^{-1}\) (corresponding to 2.0 m s\(^{-1}\)) slightly to the relative west of the center at approximately 800 hPa. The ascent slowly decelerates thereafter and the bilobal shape becomes asymmetric, favoring the relative west side, coming almost to a stop by the tropopause. Figure 10 also shows the zonal circulation through the center of the composite. Perhaps most apparent is the strong blocking signal at most heights from the Himalayas in the east—a considerable majority of the wind inflow at the center comes from the west, although outflow aloft is roughly symmetric. The strong easterly inflow near the center and the surface could be the result of cyclonic flow being suppressed near the Himalayas and results in slightly easterly flow throughout the axis up to the 350-hPa level (potentially supporting the westward axial tilt). Divergence (not shown) is rather noisy, although there is an area of concentrated, strong convergence at the center near the surface; near the tropopause, where the outflow is situated, there is an area of weak divergence (approximately 5 times weaker than the convergence at the surface), slightly to the relative west of the center.

d. Moisture flux

For horizontal moisture transport in the composite, we must consider sources of both water and advection and do so by separating the specific humidity flux vector into contributions from both the mean state and the depression-induced perturbation; that is,

\[
q\mathbf{v} = \overline{q}\mathbf{v} + q'\mathbf{v} + \overline{q}' + q'\mathbf{v}',
\] (1)
where $q$ is specific humidity, $\mathbf{v}$ is the wind vector, the overbar represents the mean state of the variable (i.e., the climatology), and the prime represents the depression anomaly to the mean state (i.e., the perturbation). The presence of a depression contributes only a fraction of the total moisture present in the composite (not shown), and the base monsoon state contributes little to the depression winds near the center (Fig. 4); therefore, we might expect the most significant contribution on the right-hand side of Eq. (1) to come from the anomalous (i.e., depression induced) advection of climatological moisture ($\overline{qv}$). To confirm this, we examine the structure of each perturbation term (shown in Fig. 11). We see that the greatest contribution indeed comes from $\overline{qv}$ and that, although the contribution from $q'v'$ is roughly a fifth as large, it has greater relative vertical and horizontal extent and a more strongly bifurcated structure.

e. Temperature profile

The vertical structure of temperature is given in Fig. 12a and is shown (owing to the strong temperature gradient with height) as an anomaly relative to the structure of the boreal summer mean temperature profile. The resulting composite is a cold-core event in the lower troposphere, agreeing with Sarker and Choudhary (1988) and Hurley and Boos (2015), amongst others. Sarker and Choudhary (1988) suggest the reasons for the cold anomaly include dynamical uplift, evaporating precipitation, and radiative cooling. We speculate that Fig. 8 corroborates the last point, indicating that cloud cover reduces insolation (cooling the lower levels) and blocks outgoing longwave radiation (heating the upper levels)—a mechanism that would be absent if there were the fully developed eye of a tropical cyclone. The warm core aloft is a result of the high instability at lower levels and resulting rapid ascent, coupled with strong latent heating from the high rates of convective precipitation associated with depressions, although its magnitude ($\sim 2 \text{ K}$) is considerably less than that of a tropical cyclone (La Seur and Hawkins 1963; Hawkins and Rubsam 1968; Hawkins and Imbembo 1976; Stern and Nolan 2012).

For comparison, the potential temperature anomaly is shown in Fig. 12b. We see very little difference in the lower levels (i.e., before much lifting has taken place), although a warmer core (+2, 0.4 K warmer than temperature alone) is present in the midtroposphere, strongly agreeing in shape and magnitude with the composite from Hurley and Boos (2015). The vertical gradient implies that, relative to the climatology, latent heating has the greater effect until 350 hPa while radiative cooling has the greater effect thereafter until the tropopause. The equivalent potential temperature (Fig. 12c) shows the importance of moisture in the depression structure—the maximum has lower altitude (around 650 hPa) and larger positive anomaly ($\sim 5.8 \text{ K}$).

We categorize the 106 depressions into warm core, neural core, or cold core by calculating the average 1000-hPa temperature anomaly in a 1.5° box surrounding the center. Using $T_{\text{warm}} \geq 0.75 \text{ K} > T_{\text{neutral}} \geq -0.75 \text{ K} > T_{\text{cold}}$, we count 2, 28, and 76 events, respectively. While SSTs are usually used in the description of tropical cyclones, the significant duration spent by monsoon depressions over land meant that surface air (1000 hPa) temperature provided a more useful measure. The criteria were established by identifying and removing the two warm-core outliers, then fitting a two-Gaussian sum (giving a better
mean-square error than 1) to the distribution of central temperatures, constraining one to be centered on the origin. The strongest cold-core event registered at 3.6 K (with a mean of 1.9 K). Neutral-core events exhibit a much steeper temperature anomaly gradient approaching the tropopause and larger anomalous heating aloft (Fig. 13). The sample of warm-core depressions is too small to draw any significant conclusions. The neutral-core-minus-cold-core PV composite (not shown) has lower tropopause height above cold-core depressions and higher PV in the flanks near the surface; the centers of each type appear very similar. Also not shown graphically is the comparison of surface rainfall rates between cold- and neutral-core events; these are given in Table 1. The greater rainfall rates, and the slightly larger area of influence of neutral-core events, should be expected upon inspection of the temperature profiles, which show that neutral-core events exhibit a warmer core aloft, indicating stronger latent heating.

f. CAPE and tephigram analysis

Variables relating to convection are important because they provide an insight into how thermal energy is distributed and used within a system. Figure 14 shows that the maximum CAPE is found approximately 5° ahead of the center, reaching a peak value of 2050 J kg\(^{-1}\). The minimum CIN is found near the center, with a value of −1.2 J kg\(^{-1}\) but increasing exponentially ahead of the depression. There are two competing effects at the center: a strong positive moisture anomaly (reducing the CIN; i.e., less negative) and a steeper temperature anomaly gradient (increasing the CIN); however, the moisture anomaly has a much stronger effect and dominates the CIN.

Figure 15 shows soundings through the center (single grid point) in the composite depression and two subsets. The full composite (Fig. 15a) has evidence of strong ascent through the lower and middle troposphere, and we can deduce large CAPE by considering the saturated pseudoadiabat coinciding with Normand’s point (approximately 950 hPa). Extrapolation of this adiabat to the upper troposphere suggests a level of neutral buoyancy at approximately 120 hPa. We also split the composite into two pairs of subsets: day–night (Fig. 15b) and land–ocean (Fig. 15c). The day–night tephigram indicates that the daytime sounding is slightly warmer and drier throughout its height, though the more substantial temperature and moisture differences in the boundary
layer engender high CAPE and very small CIN during nighttime (lower temperature but same dewpoint raises the relative humidity). The land–ocean contrast shows that, particularly in the midtroposphere, there is generally a cooler, moister environment over the ocean with lower CAPE and greater CIN.

Calculable products from these tephigrams and several others (not shown) are given in Table 2. Important results are larger CAPE and reduced (less negative) CIN in the final day (almost always over land) when compared to the first day (almost always over ocean) and larger CAPE in neutral-core events than cold-core events [because, despite the shorter distance (in pressure units) between the level of free convection and the level of neutral buoyancy, the parcel temperature is comparatively much higher—compare Fig. 13a and Fig. 13b]. In neutral-core depressions CIN is nearly 3 times greater than in cold-core depressions owing to higher surface temperature and resultant lower relative humidity. This is a significant difference because there is very low variance in CIN at the center of the composite (not shown).

In general, the level of neutral buoyancy is near the top of the cloud deck (cf. Fig. 8c), and any difference is likely accountable to entrainment of surrounding cool, dry air. The lifted condensation level is comparable to the bottom of the cloud deck.

We examine variations in profiles according to the diurnal cycle or surface type in the next section.

g. Summary of composite vertical structure

Figure 16 shows some key features of the vertical structure presented throughout this section. In particular, we note the generally asymmetric structure: a westward axial tilt in many fields and an intensification of the lower-tropospheric winds on the Himalayan side of the vortex.

5. Variability within the composite

Although most of our composite diagnostics have been presented with a test of statistical significance, we also discuss the stability of the composite, first using standard deviation of the entire composite, and second by breaking the composite down and contrasting land, sea, and transitional (coastal) phases. The latter is particularly important as the conditions surrounding a depression spinning up in the Bay of Bengal are very different to those as it decays over north India. We also consider the diurnal contrast within depressions.

a. Standard deviation

Most fields are highly homoscedastic, with a standard deviation far smaller than the anomaly to the climatology;
hence, the large areas of significance in the horizontal and vertical cross sections presented above. Variance invariably reaches a maximum in the deep flanks of the composite, where influence from the major processes within the depression is very low, and the effects of varying environment, composite rotation, and different depression sizes start to become significant.

b. Variability over the lifetime

If we can understand how the structure of depressions transforms over time, we can better represent their processes in models and better predict their durations and trajectories. Almost all depressions form over the ocean and eventually die over land, so there is likely to be some correlation between young (old) depressions and ocean (land) depressions: the latter is discussed in detail in the next section.

Figure 17a shows how the central pressure, wind, and vorticity extrema vary as a function of depression age. The variables for each depression were interpolated onto a lifetime-percentage array, rather than using an absolute time scale because depressions have considerably variable durations (see Fig. 17b). All variables demonstrate a midlife extremum: central pressure (represented as geopotential height anomaly) reaches a minimum before wind speed and relative vorticity, both of which have maxima around 60% before decaying significantly and rapidly during the dissipation phase of the depression. If we compare first-day and final-day PV composites (not shown, but see Fig. 9a), the most immediate difference is the extent of the upper anticyclone, which has a larger (more positive) PV in the flanks in the final day when compared to the first one. Differences near the surface extend to approximately 550 hPa and feature generally lower (less positive) PV in the center and higher PV in the flanks on the final day.

*TABLE 1.* Some data showing rainfall rates (mm day$^{-1}$) of selected percentiles for cold- and neutral-core depressions over the domain used in, for example, Fig. 14.

<table>
<thead>
<tr>
<th>Percentile</th>
<th>Cold core (mm day$^{-1}$)</th>
<th>Neutral core (mm day$^{-1}$)</th>
</tr>
</thead>
<tbody>
<tr>
<td>50</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>75</td>
<td>1.75</td>
<td>3.47</td>
</tr>
<tr>
<td>90</td>
<td>22.9</td>
<td>29.0</td>
</tr>
<tr>
<td>95</td>
<td>50.4</td>
<td>60.2</td>
</tr>
<tr>
<td>99</td>
<td>141</td>
<td>165</td>
</tr>
<tr>
<td>99.9</td>
<td>305</td>
<td>339</td>
</tr>
</tbody>
</table>
A histogram giving the lifetimes of each individual depression tracked is shown in Fig. 17b. It shows the shortest-lived depressions last approximately a day: this is the threshold for the IMD and our tracking software, but depressions can have lifetimes as long as nearly 10 days, with a mode of just over 3 days. This distribution is the reason information in Fig. 17a is presented as it is. Several well-known probability distributions were fitted to the data in Fig. 17b, with the generalized logistic distribution (shown) having the smallest mean-square error.

We also find that there is no significant relationship between depression frequency and time of year and that the frequency has little effect on all-India rainfall; Raghavan (1967), Dhar and Rakhecha (1976), and Krishnamurthy and Ajayamohan (2010) showed that this is because low pressure systems result in excess rainfall in some areas but deficient rainfall in others. The correlation coefficient between trough rainfall and the depression frequency for the relevant period (1998–2013) is 0.77.

c. Coastal transitions and the land–sea contrast

Land–atmosphere and ocean–atmosphere interactions can differ from each other considerably, so we expect monsoon depressions (which as we have seen, are fairly dependent on surface interactions) to differ considerably also. To explore the effect of surface type (land, coast, ocean) of monsoon depressions, we first define a coastal transition region, sufficiently wide to span the typical horizontal scale of the core of a passing depression, but not so large as to prejudice the sample sizes of land or ocean composites. A threshold of ±200 km either side of the coast was chosen, providing a sample of 680, 687, and 159 6-hourly depression time steps for land, coastal, and ocean areas, respectively, with the advantage that many fields (e.g., potential vorticity, geopotential height anomaly, vertical wind speed) have values of less than half their central extrema at this radius. The location of the center is used to define the entire depression.

For specific humidity (not shown), the difference between the three partial composites is again significantly less than the standard deviation (by a factor of 4) across the full composite, even though this variable is so dependent on surface composition. There is, however, a “tightening up” in the lower layers, where the horizontal extent of the central maximum at each height is reduced at each progressive stage. This is a common theme throughout the comparison of ocean–land composites—the magnitude of any central extremum differs insignificantly if at all, but in many fields we observe an increasing confinement to the central axis, resulting in a steeper horizontal gradient. The same analysis with potential vorticity (not shown) reveals little significant variation across the surface types.

There is little difference between the three partial composites for equivalent potential temperature anomaly (Fig. 18): the maximum is highest in the land-based composite but does not differ much from either the ocean- or coast-based composites. All three demonstrate a warm core at approximately 650 hPa.

CAPE around the center of the land-based composite is much higher than its ocean-based counterpart (Table 2), but this maximum has a much smaller extent (~6’ wide as opposed to ~20’ wide; not shown).

We also separately composited points according to their instantaneous location, resulting in “purely oceanic” and “purely land-based” composites. Figure 19 shows the wind structure of such a composite: average sample sizes are 465 time steps for ocean and 1060 for land in this and subsequent composites. The near-surface winds around the core are weaker over the land, partly as a result of friction, and partly as a result of the loss of warm water vapor as an energy source. The strong winds also become a lot more constrained to the center as the depression passes over land and weakens. The difference between the two in Fig. 19c reflects the weakening and constriction of the winds near the surface. The tropospheric winds round the core strengthen more to the relative east than to the relative west, possibly owing to orographic channeling by the Himalayas.

Further ocean-minus-land vertical composites (as Fig. 19c) are shown in Fig. 20. Specific humidity shows little contrast between the two in the lower levels, with
the only significant difference being the drier flanks above the ocean at a radius of $\sim10^\circ$. In the middle and upper troposphere, however, the ocean composite has considerably more moisture, particularly outside the center. The difference in relative vorticity (Fig. 20b) shows that the composite is less constrained to the center over the ocean and that the vortical core penetrates nearer to the ocean surface than to the land, a plausible result of orographic friction. So the depression is larger and more intense over water than land, although these differences are fairly small when compared to the mean structure. The tropical summer atmosphere is generally cooler over the ocean than land, so it is important to see how the depression organizes this difference, shown in Fig. 20c. As expected, it is cooler throughout, although the deep latent heating from increased precipitation removes much of the difference at and near the center; in the flanks, the difference is of order 1–2 K, and explains much of the counterintuitive structure of the difference in specific humidity. Surface latent heat flux (not shown) is substantially larger over the ocean (maximum 185, average 150 W m$^{-2}$) than over land (maximum 95, average 65 W m$^{-2}$): as expected, the maxima are collocated with lowertropospheric wind speed and they have local minima at the center. Surface sensible heat flux (not shown) also has central minima over both land and ocean, following the surface temperature structure. As in the full composite, there exists a boundary layer height minimum of equal magnitude ($\sim500$ m) over both the land and ocean composite centers, rising to the respective climatological values within several hundred kilometers.

![Figure 15](image-url)

**Fig. 15.** Tephigrams of simulated soundings going through the center of the composite depression and selected subsets. Temperature in red; dewpoint temperature in blue.
d. Diurnal cycle

We have already noted contrasts between composite depression structure in day–night conditions. As ERA-Interim data are released in 6-hourly time steps starting with 0000 UTC, the night composite was constructed from 1800 and 0000 UTC (2330 and 0530 Indian standard time), and the day composite from 0600 and 1200 UTC. We examined the day–night difference in selected composite variables (Fig. 21); only a few of many fields examined showed significant differences away from the center. Figure 21a shows that the day composite is almost ubiquitously 0.2 K hotter, with the increase rising to over 1.3 K in the lower troposphere, this maximum being due to strong thermal interaction between the atmospheric boundary layer and the surface (see also section 4f). Figure 21b shows that (as in Fig. 15b) the daytime depression is also drier throughout in terms of relative humidity. Figure 21c shows that there is a nearly universal reduction in cloud cover by day, particularly in the low-level cloud at the center.

6. The influence of large-scale forcing

We now consider whether organized modes of variability such as intraseasonal and interannual modes connected to the monsoon can influence the structure of monsoon depressions.

a. El Niño–Southern Oscillation

El Niño–Southern Oscillation (ENSO) is responsible for a significant fraction of the interannual variability within the monsoon system, affecting precipitation by as much as 20% (Webster et al. 1998). It is therefore plausible that synoptic-scale systems (such as depressions) within the monsoon might also be significantly affected by ENSO. We define an ENSO event as a contiguous seven month period with a Niño-3.4 region index of greater than 0.5 K. Such an episode is then linked to the monsoon season that precedes it.

While warm-core tropical cyclone activity is enhanced over the Indian Ocean during La Niña and suppressed
during El Niño (e.g., Bell et al. 2014), we find (Table 3) that the reverse is true for cold-core depressions, with approximately 16% more activity during El Niño years than La Niña [cf. 12% in the Hurley and Boos (2015) dataset]. Krishnamurthy and Ajayamohan (2010) studied monsoon low pressure systems (rather than specifically depressions) and found a slight increase in low pressure system (LPS) days during La Niña.

Figure 22a shows the El Niño-minus-La Niña rainfall composite for depressions in La Niña years during the TRMM dataset period. The overall rainfall rate, away from the center, in La Niña depressions is considerably

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**FIG. 17.** (a) Surface pressure anomaly at center (hPa), maximum 850-hPa wind speed anomaly (m s⁻¹), and 850-hPa central relative vorticity anomaly (10⁻⁵ s⁻¹) of the composite depression as a function of percentage of lifetime duration, with respective percentage of systems over land, coast (defined as within 200 km of coastline), and sea shown underneath, anomalies are relative to a 21-day running mean. (b) Histogram of depression durations, with the generalized logistic function fitted. The mean depression duration is 3.3 days, with a standard deviation of 1.6 days.

**FIG. 18.** Vertical structure of equivalent potential temperature (K): (a) ocean, (b) coastal, and (c) land composites. Colors are grayed out where the subset is not significantly different from the full composite at the 95% confidence level.
greater than that of El Niño years; this agrees with the strong correlation between drought–flood years and El Niño–La Niña years, respectively (e.g., Rasmusson and Carpenter 1983; Ropelewski and Halpert 1987; Webster and Yang 1992). However, there is evidence of a stronger, more moist core to the El Niño depressions.

ERA-Interim reanalysis data in time steps of 6 h show a total of 381.5 depression days spread among the 106 depressions (Table 3). For El Niño years, there were 7 and 29 neutral- and cold-core depressions, respectively, whereas for La Niña years, the count was 10 and 23 (with 11 and 22 for neutral years). So, for neutral and La Niña years, cold-core depressions were only a little over twice as likely as neutral-core counterparts; whereas for El Niño years, they were over 4 times as numerous here. This could be in part due to El Niño depressions being drier (Fig. 23c), allowing greater reduction in temperature via latent heat of evaporation. However, the sample sizes are small, so the results may not be significant.

Figures 23a–c show the vertical structure for several variables in the El Niño-minus-La Niña composite. The structure of the geopotential height (see Fig. 23a) is quite different. We see the surface is almost entirely at a higher geopotential height (i.e., higher pressure) for the El Niño composite compared to the La Niña one. With altitude this falls, with a 0-m isohypse passing over the center at approximately 420 hPa. Therefore, an El Niño depression is typically colder than a La Niña depression from the midtroposphere upward (Fig. 23b), since pressure falls off more rapidly with height. This is probably because weaker mean-state convection in El Niño depressions releases less latent heat aloft. The differences in structure of relative vorticity (not shown) indicate the core is considerably less cyclonic in El Niño depressions, remaining so (with reducing intensity) up to the tropopause. This comparatively less vigorous core is surrounded by zones of higher relative vorticity that also persist and spread out aloft, indicating that these depressions have a less tightly wound center but stronger rotation in the flanks.

As we might expect from Fig. 22a, the La Niña composite (Fig. 23c) is generally relatively moister, especially in the midtroposphere, with some closed $-6 \times 10^{-4} \text{kg kg}^{-2}$ isohumes around 600 hPa in El Niño relative to La Niña. There is also a surface gradient, with the El Niño composite possessing comparatively drier air in the relative west. A more detailed analysis of the ENSO differences is beyond the scope of this study.

b. The Indian Ocean dipole

The Indian Ocean dipole (IOD) was first identified by Suji et al. (1999) and we represent it by their dipole mode index (DMI), defined as the difference in SST...
anomalies between two boxes in the west (10°S–10°N, 50°–70°E) and southeast (10°S–0°, 90°–110°E) Indian Ocean. Values greater than 0.5 (less than −0.5) are named positive (negative) IOD events. Krishnan et al. (2011) considered the positive IOD event of 2006 and its contemporaneous depressions, showing that there was a positive feedback system that contributed to the longevity of depressions. They subsequently generalized their study to a climatology and found that positive IOD events added approximately 12% to the lifetime of depressions. Here, we briefly explore (see Figs. 23d–f) the difference in structure between monsoon depressions in positive and negative IOD environments. The state of the IOD has a much lesser impact on depressions than does ENSO.

The difference in geopotential height is shown in Fig. 23d and indicates that depressions associated with positive IOD events are slightly weaker and larger. This is corroborated by the temperature structure (Fig. 23e), which shows a weakening of the warm core aloft. This may be connected to the cooling of the ocean at the head of the Bay of Bengal, where most depressions originate (Fig. 24a), resulting in less moisture being available from evaporation for early intensification processes. However, as Krishnan et al. (2011) showed, positive IOD events result in greater cross-equatorial moisture flux and an increase in the meridional shear of the zonal wind (resulting in greater barotropic instability), which supports the depression in its transit across the trough and stalls the dissipation. The difference in specific humidity between positive and negative IOD events (Fig. 23f) displays little coherent structure and is not significantly different from zero almost anywhere.

The state of the subcontinent in both 850-hPa winds and surface temperature is given for IOD positive-minus-negative and ENSO positive-minus-negative depression days in Figs. 24a and 24b, respectively. The well-defined IOD structure in SST induces an anomalous north–south land surface temperature gradient and a slight strengthening of the monsoonal winds. In contrast, the ENSO difference pattern shows a slight weakening of the monsoonal winds and a warmer surface owing to deficient rainfall. To conclude, El Niño conditions may produce more depressions than La Niña, but they are weaker, whereas the IOD makes no significant difference to the structure of depressions. The influence of the IOD on surface rainfall (not shown) is also slight, but in agreement with the small difference in temperature: the slight drop in temperature throughout the profile of the IOD positive composite is accompanied by a small reduction in precipitation. The maxima in both composites are still found in the relative southwest sector, with an IOD

Fig. 20. Vertical structure of the ocean-minus-land composite (as in Fig. 19c) for specific humidity (10⁻⁴ kg kg⁻¹), relative vorticity (10⁻⁵ s⁻¹), and temperature (K). Significance as in Fig. 13c.

(a) Specific humidity  
(b) Relative vorticity  
(c) Temperature
positive maximum of 50 mm h\(^{-1}\) and an IOD negative maximum of 56 mm h\(^{-1}\).

c. Active and break phases

Active (enhanced rainfall) and break (suppressed rainfall) phases of the monsoon significantly affect both spatial rainfall distribution and depression frequency (Krishnamurthy and Shukla 2007), with far more depression days during active periods. We can combine ENSO-based prevalence with active–break prevalence (see Table 3). We used data on active and break spells from Pai et al. (2014), requiring constriction of the dataset to July and August 1979–2010 (leaving 51% of samples).

Because of the relative sparsity of depressions coinciding with breaks (Krishnamurthy and Shukla 2007, also Table 3), meaningful statistics of break depressions were unrealizable. Therefore, we compare depression composites in active and neutral spells (see Figs. 23g–i). Active-spell lower-tropospheric geopotential height (Fig. 23g) is much lower pressure in the center, with a much steeper vertical gradient, suggesting that depressions intensify under active conditions.

The maximum magnitude of the temperature difference is little greater than 0.5 K (Fig. 23h), but active depressions clearly exhibit a slightly colder core at the surface and a warmer upper troposphere because of the greater release of latent heat through the stronger convection. This is an amplification of the structure seen in Fig. 12a and further implies a strengthening of depressions under active monsoon conditions.

The active composite has a much more cyclonic core, flanked on either side by areas of considerable relative anticyclonicity, and with a weaker corresponding anticyclone. The composite active depression is considerably wetter in the midlevels (750–500 hPa) as expected (Fig. 23i). This difference composite also exhibits a slight relative east–west gradient at the surface (as with El Niño–La Niña): wetter in the west and drier in the east.

The average relative vorticity (again, as an anomaly to the climatology) over South Asia for depression days is illustrated in Fig. 25a. The most immediately apparent

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**TABLE 3.** Distribution of depression days with respect to ENSO and active–break cycles.

<table>
<thead>
<tr>
<th></th>
<th>Active</th>
<th>Neutral</th>
<th>Break</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>El Niño</td>
<td>39</td>
<td>90.75</td>
<td>0</td>
<td>129.75</td>
</tr>
<tr>
<td>Neutral</td>
<td>20.25</td>
<td>97.25</td>
<td>1.75</td>
<td>119.25</td>
</tr>
<tr>
<td>La Niña</td>
<td>27.25</td>
<td>104</td>
<td>1.25</td>
<td>132.5</td>
</tr>
<tr>
<td>Total</td>
<td>86.5</td>
<td>292</td>
<td>3</td>
<td>381.5</td>
</tr>
</tbody>
</table>
feature is the strong anticyclonic vorticity band located across the Himalayan foothills, seen already in Figs. 5a and 9c. The corresponding region of anomalous positive relative vorticity is strongly correlated with the track density of depressions (see Fig. 1a), with a collocated maximum on the coast of Odisha and West Bengal. There are also several smaller, weaker areas of negative anomalous relative vorticity, notably over Pakistan where many lows eventually dissipate (Krishnamurthy and Ajayamohan 2010). This pattern is in agreement with Godbole (1977). Figure 25a also shows the depression day winds (at 850 hPa) less the climatology, revealing a structure very similar to that of an active-minus-break composite (Webster et al. 1998), and bearing some resemblance to the equivalent ENSO pattern (Fig. 24b), but it seems to bear no relation to the strength of the monsoon itself (Turner et al. 2005); this demonstrates that certain aspects of large-scale variability (La Niña and active phase) favor the circulation anomaly in which depression days occur. The sign change in zonal wind that occurs at \( \sim 20^\circ N \) is greatly assisted by the addition of cyclonic vorticity from the presence of depressions in the north of the Bay of Bengal and over northeast India. The shear imparted on the 850-hPa easterly by the Himalayas is also visible. For comparison, the difference in 850-hPa winds and relative vorticity between active depression days and nonactive depression days is shown in Fig. 25b. It demonstrates that the active state is an amplification of both the monsoon and the depression-day anomaly, and it therefore corroborates what we saw in Figs. 23g-i, that depressions occurring when the monsoon is active are somewhat more intense (lower central pressure, greater vertical temperature gradient, more moisture) than when it is in a normal or break phase.

7. Conclusions

Monsoon depressions are cyclonic tropical lows, averaging 2–5 days in lifetime, that spend much of their time over the Indian subcontinent. They are important as they contribute much rain to the northeast of the Indian peninsula, a region known to be particularly dry in coupled GCMs (Sperber et al. 2013). It is therefore important to quantify their structure, movement, and behavior, so as to evaluate and understand their dynamics. This will in due course enable evaluation and improvement of numerical models, leading to deeper understanding and to more reliable prediction, and thus to improved local community preparedness for flooding and related impacts.

We confirmed the existence of a rainfall maximum in the southwest sector of a rotated northward-propagating depression (Ramanathan and Ramakrishnan 1933; Mooley 1973; Godbole 1977). There are several orders of magnitude between the rainfall maximum of the composite (38.6 mm day\(^{-1}\)) and the maximum individual event (1.6 m day\(^{-1}\)), indicating the diversity and local detail of such extreme events.

Reanalysis data were used on the 106-depression composite to determine horizontal and vertical structure,
Figure 23. Vertical structure of the (a)–(c) El-Niño-minus-La-Niña, (d)–(f) IOD positive-minus-negative, and (g)–(i) active-minus-neutral composites for (a), (d), (g) geopotential height (m); (b), (e), (h) temperature (K); and (c), (f), (i) specific humidity (10^{-2} kg kg^{-1}). The colors are grayed out in areas where the composite difference does not significantly differ from zero. Magenta line contours represent the same variables with 21-day averages subtracted—this shows the composite difference left once large-scale forcings have been removed.
showing that many fields have their maximum intensity in the relative southwest sector. The Himalayas were found to have a pronounced anticyclonic effect on the relative east-northeast sector of the depressions, both at the surface and aloft, as confirmed by the relative vorticity field during depression days, which exhibited an intense band of anticyclonicity along the foothills. A schematic diagram of the major results is shown in Fig. 16. We confirmed the warm-over-cold temperature core structure is present in almost all depressions, relating the near-surface cold core to reduced insolation from increased cloud cover and the warm core aloft to strong latent heating from convective precipitation. We also found a bimodal potential vorticity structure at the center [also recently discovered by Hurley and Boos (2015)], as well as an intensification of the tangential winds on the Himalayan side of the depression.

The slight westward axial tilt observed in, for example, Godbole (1977) was also confirmed in the lower levels in many fields, though some fields also exhibited an eastward axial tilt near the tropopause.

We have shown that depressions can be separated into three separate core types: warm (rare and tropical cyclone–like), neutral (weak temperature anomalies near
the surface, strong warm core aloft, strong temperature anomaly gradient near the tropopause), and cold (much colder at the surface, warm core aloft, most frequent type); we found that cold-core depressions are comparatively more likely to form in El Niño conditions than neutral or La Niña ones and that neutral-core depressions are responsible for greater and more widespread rainfall.

We compared composites of depressions over land and ocean. Most differences were not large given the scatter between depressions, but depressions over land had weaker near-surface winds owing to friction and more anticyclonic vorticity to the relative east owing to the Himalayas. There was no significant difference in potential vorticity between the subsets, although equivalent potential temperature had a more extended maximum (and at a lower altitude) over land owing to differing moisture structures. We also found that the boundary layer height drops over the center of the depression. Latent and sensible surface heat fluxes were also considered as both have minima at the center and strong land–ocean contrasts, latent heat flux being much higher over the ocean.

Depressions during La Niña have moister, more vortical cores than during El Niño. They also have a warmer middle troposphere and slacker vertical geopotential height gradient, and though they yield less intense precipitation at the center, they are associated with higher rainfall rates in the flanks. The influence of the Indian Ocean dipole on monsoon depressions was very slight.

Depressions occurring during active monsoon phases are more intense: greater lower/midtropospheric geopotential height anomaly, greater vertical temperature anomaly gradient, and substantially more moisture throughout. The active-period intensification of the background monsoon conditions strongly resembles an amplification of the depression-day anomaly relative to the monsoon climatology.

Monsoon depressions are colder, wetter, and have more cloud throughout during the local nighttime than the daytime.

Our diagnostics represent a firm basis for assessing monsoon depression representation and behavior in NWP and high-resolution climate models.

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