

1 **An indirect, negative radiative effect of water vapor in**
2 **the tropics and its implications for regional surface**
3 **heat stress**

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8 **Key Points:**

- 9 • An indirect, water vapor related surface cooling is observed at the daily scale, pan-
10 tropics, bordering the ITCZ, throughout the year
11 • Lower tropospheric humidity is negatively related with local cloud radiative forc-
12 ing in these regions dominated by mid level clouds
13 • Regions occur along the horizontal branch of Hadley cell, move with the seasonal
14 cycle, possibly consequential for pre-monsoon heat stress

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Abstract

We report an indirect, negative, cloud-mediated, surface radiative effect (RE) of water vapor (IWVE) in certain regions in the tropics, which may be consequential for day-to-day regional heat stress. Using reanalysis and satellite data we show that this effect is marked by a surprisingly dominant positive relationship of cloud RE with near surface and column humidity. These clouds are predominantly low level and altocumuli, previously reported to have a negative surface RE, possibly lending the net negative RE to water vapor. Also reported earlier, these clouds form in the mid-troposphere, as detrainment offshoots of deep convective towers and can be advected away to large distances, hence requiring no local convective triggering in the IWVE regions. Evidently, the IWVE are co-located with the horizontal branch of the Hadley cell, with the lowest vertical forcing in the tropics. Moreover, these are also the transition regions between the highly cloudy and the driest parts of the tropics, with a waning down occurrence of cirrus, deep convective and altostratus clouds, linked with positive RE, corroborating the hypothesis. IWVE regions also show a large temporal variability in humidity possibly providing opportunity for a large variability in cloud fractional coverage, however the mechanism controlling this covariability is not understood. The IWVE is tightly tied with the seasonal cycle of the ITCZ and hence is likely a dominant source of pre-monsoon surface temperature variability and heat stress in the current climate. The evolution of the IWVE under future climate warming needs further investigation.

Plain Language Summary

The greenhouse radiative effect of water vapor on surface temperatures is widely known. However, can water vapor cause an ‘apparent’ radiative cooling effect under certain conditions? Here, we show that in certain extensive tropical regions, bordering the ITCZ, the lower tropospheric water vapor has a substantial relation with cloud occurrence, which subsequently renders an indirect surface cooling effect in high humidity conditions. These clouds are majorly shallow and mid level cumuli, which are known to have a negative surface radiative forcing. What causes the positive relation between lower tropospheric humidity and these clouds, is not understood. However, that these regions border the ITCZ and hence are co-located with the horizontal branch of the Hadley cell, suggests low amounts of convective triggering and hence high level clouds. The potentially reduced warming from the high level clouds may give the lower level clouds an upper hand in the net radiative impact. Lastly, these regions precede the ITCZ in time as they move north-south with it. This may be consequential for pre-monsoonal heat stress in tropical regions.

1 Introduction

Water vapor is frequently dubbed as the most important greenhouse gas which plays a dominant role in climate change through the water vapor feedback (Soden et al., 2002, 2005; Forster et al., 2021; Dessler et al., 2013). This control is delivered through the dominant impact of the increase in the upper tropospheric humidity on trapping the outgoing longwave radiation (*OLR*) and recycling it within the climate system (Soden et al., 2005; Allan et al., 1999; Dessler et al., 2013). This trapping of energy warms up all components of the climate system - the atmosphere, land and oceans. However, on hourly to daily time scales, it is the near surface humidity (or column integrated water vapor, as they are strongly correlated) which dominates the greenhouse effect of water vapor as manifested at the land surface (Shakespeare & Roderick, 2021; Flerchinger et al., 2009; M. Li et al., 2018; Brutsaert, 1975; Carmona et al., 2014). Idealized modeling studies of clear sky surface downwelling longwave radiation (*DLR*) are able to explain about 97% of the temporal variability in *DLR* with near surface humidity (q_{sfc}) and surface temperatures (T_{sfc}) (Shakespeare & Roderick, 2021). High positive correlations have been

65 reported between in-situ observed q_{sfc} and DLR (or T_{sfc}), at daily to monthly time scales,
 66 in all sky conditions (Gaffen et al., 1992; Ross et al., 2002; M. Li et al., 2018). However,
 67 exceptions to this behavior (Ross et al., 2002), relatively understudied, highlight the role
 68 of the several other complex climatic processes that water vapor involves itself in - sur-
 69 face evapotranspiration, cloud formation, convective amplification etc - in impacting T_{sfc} .
 70 The resulting processes may modulate the initial water vapor greenhouse effect at re-
 71 gional scales. We set out to investigate this regionality in the ‘effective’ water vapor green-
 72 house effect, particularly in the tropics, and investigate the role of clouds in causing the
 73 exceptional behavior.

74 This study is motivated by the need to understand the regionality in the q_{sfc} - T_{sfc}
 75 relationship, to better analyse the recent increase in global heat stress (Alizadeh et al.,
 76 2022; Pai et al., 2013; Mazdiyasn et al., 2017; X. X. Li, 2020; D. Li et al., 2020), which
 77 is predicted to increase to deadly levels in tropical latitudes under climate change, es-
 78 pecially in transition regions like the Sahel, the Cerrado and southern Asia (Coffel et al.,
 79 2018; Mahlstein et al., 2011; Coumou et al., 2013; Hoegh-Guldberg et al., 2018) affect-
 80 ing low income communities and ecosystems of these regions (Alizadeh et al., 2022; Dunne
 81 et al., 2013). Over the Amazon-Cerrado boundary in Brazil, trees are reported to be close
 82 to their threshold temperature for photosynthesis, already impacting these ecosystems
 83 during high heat stress conditions (Araújo et al., 2021; Tiwari et al., 2021; Reich et al.,
 84 2015). Over the densely populated southern Asia, pre-monsoon heat stress has emerged
 85 as a major challenge under climate change (Rohini et al., 2016; Pai et al., 2013). These
 86 regions are additionally impacted by the combined effects of high temperature and hu-
 87 midity on perceived heat, corroborating the challenge even further (Oldenborgh et al.,
 88 2018; Wehner et al., 2016; Im et al., 2017). This surmounting evidence raises the need
 89 to understand the impact of near surface humidity on surface temperatures and the co-
 90 variability between them.

91 Over and above the impact of near surface humidity on DLR , discussed above, q_{sfc}
 92 can impact T_{sfc} through other processes that control surface warming. For example, it
 93 is suggested that evaporative cooling from intense irrigation, which will probably result
 94 in a negative correlation between the two variables, can cause a muted local warming
 95 of the surface (Oldenborgh et al., 2018; Lobell & Bonfils, 2008; Lobell et al., 2008; Puma
 96 & Cook, 2010; Douglas et al., 2009; Lawston et al., 2020). However, this process is ap-
 97 plicable to regional scales only, where a substantial amount of irrigation is practiced, which
 98 is expected only in a small part of the tropics. Additionally, this process takes effect only
 99 in the dry season of the region when precipitation is low (Biemans et al., 2016; Douglas
 100 et al., 2006). Hence, a large spatial and temporal scale impact is not expected from this
 101 mode of surface cooling, especially not in the pre-monsoon season, when agriculture is
 102 anyways low due to the low supply of riverine or ground waters for irrigation, and high
 103 heat stress conditions (Biemans et al., 2016).

104 Lower tropospheric humidity is also crucial in cloud formation and hence can po-
 105 tentially influence CRF at regional scales, a relationship which we set out to investigate
 106 in this study. As an example of the former relationship, q_{sfc} plays the lead role in con-
 107 vective triggering, supporting conditions conducive to moist convection by reducing con-
 108 vective inhibition - a higher near surface moisture has been shown to have a positive im-
 109 pact on moist convection and cumulus clouds (Findell & Eltahir, 2003a, 2003b). Free
 110 tropospheric humidity above the boundary layer is also shown to support moist convec-
 111 tion by reducing the detrainment drying of convective plumes (Derbyshire et al., 2004;
 112 de Rooy et al., 2013; Holloway et al., 2009; Schiro et al., 2016). However, such positive
 113 impact of q_{sfc} on cloud formation could be marred by other complex processes like pre-
 114 vailing atmospheric stability and horizontal advection, which may interfere with the re-
 115 lation between these variables in regions outside the tropical deep convection belt. In
 116 other words, humidity-cloud relationships may or may not be the same as the above for
 117 non-convective clouds. In this study, we show the existence of extended regions in the

tropics where q_{sfc} and CWV are dominantly, positively correlated with shallow to mid tropospheric clouds which are reported, by previous studies, to have a negative surface radiative forcing (Bourgeois et al., 2016; Bouniol et al., 2012; Riihimaki et al., 2012; Stein et al., 2011; Parker et al., 2016). These regions thereby show an ‘effective’, net negative radiative impact of q_{sfc} (CWV) on T_{sfc} on a daily scale, specifically in the pre-monsoon season when heat stress is at its highest at these latitudes.

Hence, this study focuses on understanding the impact of q_{sfc} , through its relation with local cloud cover, on daily scale surface heat stress in tropical regions during the peak, pre-monsoon, summer season. The general correlations between the humidity and surface temperature variables are presented first, subsequently relating the observation to the local cloud cover and variability. Next, the relationship of these correlations with the larger scale atmospheric dynamics is presented. Lastly, a short analysis is presented to propose a possible role of the near surface humidity in causing observed heat stress through the processes identified.

2 Methods

This study focuses on processes that impact the daily scale variability of regional surface temperatures (T_{sfc}). As such, most analysis uses daily scale data for a selected month. This choice of time scales reflects our interest in the ‘instantaneous’ response of T_{sfc} to the radiative forcing by the variability in water vapor and clouds. Since the latter process takes of the order of a few hours to manifest, the day-to-day co-variability in the variables is analyzed, if not the co-variability at sub-daily time scales. The atmospheric water vapor variability at longer time scales might itself be caused by inter-seasonal and inter-annual climate variability and climate change, but that is not the focus of this study. However, our analysis with inter-annual, monthly averaged variables results in similar inferences as the analysis with daily scale variables, providing support for the generalizability of the processes identified. Results are majorly reported from the pre-monsoon summer season of the northern hemisphere, but the inferences remain unchanged for other seasons. The month of May 2016 is chosen for some specific analysis that is limited by the size of the data set involved. 2016 was a post El Niño year transitioning to a mild La Niña, infested by high heat stress in the Indian subcontinent (Oldenborgh et al., 2018).

Surface heat stress is quantified with daily scale maximum surface temperatures (T_{max}) from reanalysis and satellite datasets. T_{max} appears frequently and takes an important role in the formal estimation of heat stress (IMD, 2002; Perkins, 2015). However, in this study, heat stress is used to refer to periods of statistically significant departures of T_{max} from the climatological or monthly mean and should not be confused with the various heat wave indices used by meteorological departments to declare periods of heat stress. Most inferences reported here with T_{max} remain unchanged for an analysis with daily averaged surface temperatures (T_{avg}) and daily maximum 2 m air temperature ($T_{2m,max}$).

The daily averaged near surface specific humidity (q_{sfc}), specific humidity at various altitudinal levels (q_z) and total column water vapor (CWV) are used to capture the radiative effect of atmospheric water vapor on surface temperatures. Previous studies have utilized these variables to show that at daily to monthly timescales, in most global land areas these variables are non-linearly but positively correlated with the surface downwelling longwave radiation (DLR) or surface temperatures (Gaffen et al., 1992; Ross et al., 2002). Although the clear sky regression coefficients between q_{sfc} and DLR are space dependent (Allan et al., 1999), there is no previous evidence of them being negative. Hence, q_{sfc} is used to represent the greenhouse impact of atmospheric water vapor on surface temperatures (Shakespeare & Roderick, 2021).

167 ERA5 is the major source of daily and monthly average meteorological fields for
 168 the years 2001 to 2020 at a horizontal resolution of $0.25^\circ \times 0.25^\circ$ (Hersbach et al., 2020,
 169 2023). Surface skin temperature and 1000 hPa specific humidity from the global anal-
 170 ysis are used to represent T_{max} and q_{sfc} respectively. Supporting analysis is performed
 171 with Level 3 standard retrievals, version 7 from AIRS (AIRS, 2019) for the period 2001
 172 to 2020 at a spatial resolution of $1^\circ \times 1^\circ$. The daily daytime surface skin temperature from
 173 the ascending branch and daily near surface specific humidity averaged between the as-
 174 cending (equator passing time - 1:30 pm LT) and descending (equator passing time - 1:30 am LT)
 175 branches are used to represent T_{max} and q_{sfc} respectively.

176 Cloud cover information is obtained from ERA5 at the same spatial resolution as
 177 that for T_{max} and q_{sfc} . Cloud fields are also obtained from CERES monthly average day-
 178 time cloud product, CldTypHist, for the period 2001 to 2020 at a spatial resolution of
 179 $1^\circ \times 1^\circ$ (Doelling et al., 2013, 2016). CERES CldTypHist characterizes clouds based on
 180 three altitude and three optical depth bins. The lowest level (1000 hPa to 680 hPa) has
 181 Cumulus, Stratocumulus and Stratus with optical depths 0.0-3.6, 3.6-23 and 23-380 re-
 182 spectively. The middle level (680 hPa to 440 hPa) has Altocumulus, Altostratus and Nim-
 183 bostratus with optical depths 0.0-3.6, 3.6-23 and 23-380 respectively. The highest level
 184 (440 hPa to 10 hPa) have Cirrus, Cirrostratus and Deep Convective clouds with optical
 185 depths 0.0-3.6, 3.6-23 and 23-380 respectively. AIRS daily total cloud cover data is also
 186 used for the same analysis.

187 Surface cloud radiative forcing (CRF) is majorly estimated with hourly scale, day-
 188 time (6:00 am to 6:00 pm LT) surface radiative fluxes from ERA5. CERES SYN1deg level
 189 3 product is also used to derive the daily scale CRF for May 2016 at a spatial resolu-
 190 tion of $1^\circ \times 1^\circ$ (Doelling et al., 2013, 2016). The CRF is estimated as a difference between
 191 the all-sky and clear-sky net radiation at the surface (Ramanathan et al., 1989).

192 Most analysis in this work is performed using statistics like linear correlation co-
 193 efficients, linear regressions coefficients and anomalies from spatio-temporal averages for
 194 daily scale data analyzed over a month. Statistical significance is obtained with a two-
 195 tailed t-test for correlation coefficients, linear regressions and differences at the 95% con-
 196 fidence level in all cases.

197 Some of the analysis is performed over the African subcontinent because the con-
 198 tiguous nature of this region across the equator into the two hemispheres provides the
 199 opportunity to investigate the processes studied along a meridian, without disruption
 200 from land-ocean boundaries and high topography. The generality of these processes, if
 201 applicable, over other continents is indicated by providing relevant details in the text.

202 Figures 1 and S1 show the multiyear mode of monthly correlations of daily scale
 203 T_{max} (or $T_{2m,max}$, T_{avg}) and q_{sfc} (or CWV). Multiyear mode is calculated using grid-
 204 scale monthly correlations for 20 years, from four neighboring grid points, regardless of
 205 the statistical significance of the correlations in individual years. However, the correla-
 206 tions of interest (IWVE correlations defined in ‘Results and Discussion’) are robustly sig-
 207 nificant across all years. Fig. S1b is shown as an example and will be discussed in de-
 208 tail in the next section.

209 3 Results and Discussion

210 The objective of this study is to identify regions where water vapor impacts sur-
 211 face heat stress through processes over and above its greenhouse warming impact. In other
 212 words, we endeavor to investigate regions where the direct greenhouse effect of water va-
 213 por may be masked by the radiative impacts of the other climatically important processes
 214 mediated by water vapor, like cloud formation. To identify such regions, a basic corre-
 215 lation analysis between daily scale surface temperature and indicators of atmospheric
 216 humidity is performed.

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3.1 A negative correlation between near surface atmospheric water vapor and maximum daily surface temperatures

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At the daily scale, T_{max} of a majority of regions around the globe are strongly positively correlated with q_{sfc} (Fig. 1), and CWV (Fig. S1 a) in ERA5 and also in AIRS (Fig. S2 a). The values of these correlation coefficients for May 2016, are displayed in Fig. S1 b to show that the spatial patterns of the multiyear mode of these correlations (Figs. 1, S1) compare well with the values for individual years. These widespread positive correlations could be attributed to the greenhouse effect of water vapor. The maximum positive correlations larger than ≈ 0.7 are obtained at high latitudes and moderate values over the tropical oceans, a feature of the surface temperature-humidity relationship reported before (Ross et al., 2002). The inhibition of these positive correlations along the inter-tropical convergence zone (ITCZ) over land could be attributable to the dominant presence of high level clouds in these regions, which may have a complex net CRF . The analysis of these regions is out of the scope of this study.

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The same figures, however, also show several tropical and sub-tropical land regions with high, statistically significant, negative correlations between daily T_{max} and q_{sfc} which occur across seasons and the two data sets used. Such signal has previously been observed over western USA using radiosonde data (Ross et al., 2002). In our study the negative correlation coefficients range between -0.36 to -0.97 for data averaged for May between 2001 and 2020 (Fig. 1). The largest negative correlations are found predominantly over tropical Africa and tropical South America throughout the year and over the Indian subcontinent, southeast Asia and northern Australia in some seasons. Similar patterns appear also with T_{avg} and $T_{2m,max}$ (Figs. S1 c and S1 d) and with inter-annual correlations using monthly averaged data (Fig. S2 b), indicating the generalizability of this relationship. This apparent water vapor related cooling is observed over land areas on the flank regions of the ITCZ and follows its seasonal cycle (Fig. 1). Hence, for example over the Indian subcontinent, it comes into prominence in the dry, pre-monsoon summer season (Fig. S3). It should be noted however that these negative correlations might maximize in a region's peak monsoon period. This is probably due to an atmospheric moistening and a sharp evaporative surface cooling due to heavy precipitation - for example in June over India when the monsoons arrive over land (Fig. S3). However, the cause of the water vapor related cooling in the pre-monsoon dry period is not obvious.

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This negative correlation of near surface water vapor with surface temperatures, suggesting an apparent negative radiative forcing from water vapor is termed here as the 'Indirect Water Vapor Effect' or IWVE. This apparent indirect forcing of water vapor is investigated next. In the following, IWVE correlations will mean a correlation between daily q_{sfc} and T_{max} , unless otherwise specified. Furthermore, the positive IWVE correlation regions over the deep convective regions near the equator are termed as the ITCZ and the positive IWVE correlation regions polewards of the IWVE regions are termed as the desert or DES regions.

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Figure 2 shows the correlations between specific humidity at an altitude (q_z) and daily T_{max} over three regions of the African subcontinent - an IWVE, an ITCZ and a DES region in May 2016. These regions are carefully chosen to lie at the same longitude (19°E) and away from topography. Over the IWVE region, the correlation between T_{max} and q_z is always significantly negative throughout the troposphere. Similarly, over the DES region the above correlation is significantly positive throughout. However, over the ITCZ region the correlation is positive close to the surface but changes to negative in the upper half of the troposphere. The altitudinal dependence of the correlations over the ITCZ suggests a complex interplay of multiple processes, possibly involving clouds and water vapor, impacting surface temperatures, which is out of the scope of this study. Only the IWVE regions, where q_{sfc} dominates the overall correlation between CWV and T_{max} , are studied here. This is a reasonable choice because most IWVE regions defined

based on Figures 1 and S1 also show a consistent negative correlation with CWV (Fig. 3).

3.2 IWVE regions occur between the cloudiest and clearest parts of the tropics

Figures 1, S1 and S2 also show an overlay of total cloud cover averaged over the indicated time periods from ERA5. The IWVE co-occurs with regions of transition cloud cover, tropics-wide, i.e. between regions of maximum and minimum cloud cover where its gradient is substantial. For example, during May over central Africa (Fig. 1 b), the average cloud cover over the IWVE regions is around 0.51 (standard deviation 0.17), as compared to around 0.71 (0.18) over the ITCZ and 0.24 (0.18) over the DES regions. Hence, the total cloud cover in the IWVE regions is intermediate between regions of predominant convection (ITCZ) and regions of predominant downwelling (over subtropical deserts like the Sahara, Kalahari, the Arabian desert etc). Additionally, the fact that the IWVE regions tightly border the equatorial regions of deep convection everywhere, and follow the north-south march of the ITCZ, suggests that the IWVE correlations could be mediated by the radiative effects of some specific clouds that are dominant at the borders of the ITCZ.

Next, a possible pathway, mediated by clouds, is analyzed, which may lead to the apparent negative radiative forcing from water vapor in the IWVE regions. These regions receive much lower precipitation than the ITCZ (Fig. S4) which suggests that the IWVE regions may harbor clouds, or a cloud distribution, that is distinctly different from those in the ITCZ, resulting in a distinctly different CRF . Clouds show a spectrum of surface and top of atmosphere CRF depending upon their cloud top height and optical depth (Chen et al., 2000; Hartmann et al., 1992). A radiative transfer calculation is often necessary to obtain the net radiative effect of a certain vertical distribution of clouds. However, in this study we have taken evidential support from previous research to tie the observed cloud distribution over the IWVE regions with the negative, apparently cloud-mediated radiative impact of q_{sfc} , both of which are discussed below. A comprehensive numerical analysis of these relationships is left for a follow up study.

3.3 IWVE regions are dominated by ‘cooling’ mid tropospheric clouds

Figure 4 shows the distribution of the daytime low, mid and high level clouds zonally averaged over a region in Africa (30°N , 17.5°E and 30°S , 30°E), obtained from the CERES CldTypHist product for May 2001 to 2020. The corresponding May average cloud maps for Africa are provided in Figure S5. There is a clear distinction in the distribution of different types of clouds between the ITCZ, IWVE and DES regions. The IWVE regions are marked by a dominant presence of low and mid level, optically thin to moderately thick clouds, like shallow cumuli, altocumuli and stratocumuli. While the ITCZ, apart from the low and mid level clouds, also have a higher percentage of optically thin to thick, mid and high clouds, relative to the IWVE regions. This feature is present in all seasons except when the IWVE correlations are large during the monsoon, apparently due to the presence of strong evaporative surface cooling (August in Northern Hemisphere in Figure 4c), in which case the low and altocumulus clouds become infrequent over the IWVE regions. Specifically, the non-monsoonal, IWVE regions show a dominant presence of mid level, low thickness altocumulus clouds, which follow the poleward boundaries of the IWVE regions and are present abundantly and uniformly between these latitudes (Figure S5 d).

The effective radiative impact of this disparate cloud distribution on surface temperatures over the IWVE, ITCZ and DES regions is expected to be different. Noticeably, low level shallow cumulus and mid level altocumulus and altostratus have been previously shown to have a negative radiative forcing on surface temperatures (Chen et al.,

2000; Hartmann et al., 1992). Additionally, the predominant occurrence of thin (optical depth around 1.03) mid level clouds (3 to 8 km) with a possibly negative radiative forcing over land has been previously reported in the pre-monsoon months over western Africa in the Nigerian region (Bouniol et al., 2012; Bourgeois et al., 2016; Stein et al., 2011), in Darwin (Riihimaki et al., 2012) and in the pre-monsoon Indian subcontinent (Parker et al., 2016). These mid level, thin, altocumulus clouds with a negative radiative forcing in the tropical regions (Bourgeois et al., 2016), are associated with mid level detrainment from deep convective clouds near the zero degree isotherm (Johnson et al., 1996, 1999; Iwasa et al., 2012). It is suggested that the latent heat of condensation to ice causes the deep convective tower to lose buoyancy and form these clouds, which can get horizontally advected to far away regions. As such, the occurrence and radiative impact of such mid level clouds is likely to be important in the flank regions of the equatorial deep convective belt, which is where the IWVE regions occur.

Hence, the dominant shallow- and alto- cumulus clouds in the IWVE regions could be responsible for the observed surface cooling. However, the question - how is q_{sfc} in the IWVE regions linked to surface cooling caused through these clouds? - is still open. This is an important question, because the expected positive, greenhouse radiative forcing of water vapor seems to be negated by a mechanism that supports clouds with a negative radiative impact. This mechanism should cause the aforementioned cooling clouds to occur along with high q_{sfc} in the IWVE regions. Given the strength and robustness of the IWVE correlations over several time and spatial scales, this mechanism could have great significance for the heat stress dynamics in the IWVE regions. Although the exact mechanism is not investigated here, we show evidence of such a negative relationship between CRF and q_{sfc} that exists dominantly over the IWVE regions, rendering the ‘effective’ negative radiative forcing from water vapor.

Over the IWVE regions, we find a robust, significant and large negative (positive) regression between q_{sfc} and daytime CRF (cloud fractional coverage, figure not shown), but insignificant regression between these quantities in the DES regions and positive, insignificant values in some areas of the ITCZ (Fig. 5). This feature of the IWVE, ITCZ and DES regions is observed pan-tropics (Figures 6 and S6) throughout the year (figures not shown). However, we do find the ITCZ regions to have a significant negative correlation between upper tropospheric water vapor and CRF (figure not shown) indicating a dependence of cloud formation on local environmental humidity, as is presumably also the case over the IWVE regions. In the IWVE regions, we however find this strong dependence of CRF on the water vapor present throughout the column (figure not shown). In other words, in the IWVE regions, water vapor in the lower and higher tropospheric levels seems to vary and impact cloud cover in tandem. The mechanism behind this apparent correlation is not understood, however, the correlation enables lower tropospheric water content to be a good indicator of CRF in these regions.

Lastly, the IWVE regions also show a large day-to-day variability in both cloud cover and CRF , which is not present in the DES and ITCZ regions (Fig. 5). Over DES, q_{sfc} is too low to be associated with substantial cloud occurrence, whereas, over the ITCZ it is too high to limit or control the same. While we have not established any cause-and-effect relationships between q_{sfc} and cloud cover in the three regions, the general atmospheric conditions do seem to be less conducive to or supportive of clouds over the DES regions and vice versa over the ITCZ. Some of this possibly could be attributed to the stable conditions that form over the DES regions due to Hadley circulation and vice versa over the ITCZ. The IWVE regions however show a large variability in both q_{sfc} and CRF , suggesting atmospheric conditions variable enough to support different types and amounts of clouds. This feature is observed in most IWVE regions throughout the tropics (figure not shown). The effect of tropical latitudinal overturning circulation on the IWVE regions is analyzed next.

3.4 IWVE co-located with the horizontal branch of Hadley cell - cloudiness not dominated by convective forcing

Cloud formation or occurrence is governed by several factors like water availability and the presence of a vertical lift - either due to buoyancy, low level convergence, turbulence, orography or frontal interaction. As such, a weak control from one factor can give the other processes an upper hand in impacting cloud formation or occurrence. Specifically, the absence of one factor can result in a decrease in a certain type of clouds and favor the other types. The role of a presence or absence of vertical lift on cloud occurrence in the three regions will now be investigated.

An abundant amount of convective forcing is available over the ITCZ throughout the year due to maximum solar insolation, which results in the high cloud cover and precipitation associated with these regions discussed earlier. The divergence of the horizontal wind, column integrated between 500 hPa and 200 hPa in the tropics, shows positive values (divergence) over the ITCZ regions (Fig. 7). Hence, the regions where q_{sfc} is positively (not) correlated with T_{max} (CRF) are also the regions of deep convective forcing. The high cloud cover of all types in these regions supports this inference (Fig. 4). The opposite is true for DES regions which show predominant horizontal convergence, downwelling and clear skies (Fig. 7).

The IWVE regions, on the other hand, are marked by the smallest horizontal divergence and vertical motion as depicted in Figure 7. This is seen by the overlap between the IWVE regions and the contours of near zero divergence of the horizontal winds integrated over the upper troposphere. This feature is present over all IWVE regions in the tropics. This absence of a strong vertical forcing could be related with the dominance of shallow cumulus and altocumulus clouds and the lower areal coverage of deep convective and cirrus clouds in the IWVE regions (Figure 4). It can only be speculated now and verified in a numerical study that vertical forcing is conducive to a certain type of clouds, the absence of which may be related with the IWVE.

3.5 Why could the water vapor indirect effect be important?

The co-location of the IWVE regions with transition cloud cover and negligible vertical motion (or convective forcing) indicate that these regions occur in the horizontal branches of tropical continental scale overturning circulations like the Hadley cell. Clearly, the Hadley cell is not strictly meridionally oriented everywhere, which is why we don't find a strict latitudinal dependence of the IWVE regions over various continents in a given season. In fact, continental scale overturning circulations, like the one found over the northern Sahara in May 2016 (Figure 7), can also engender IWVE regions co-located with their horizontal branches. However, Figure 1 demonstrates clearly that the IWVE is not present in regions between Libya and Mali over a climatological average and may not be important for the regional heat stress climatology. However, in regions affected by a climatological latitudinal overturning circulation, like the Hadley cell, the IWVE could be decisive in determining pre-monsoonal surface heat stress.

To this effect, we provide an example of the surface heat stress during the severe, pan-India heat wave of May 2016. An exhaustive analysis of the general heat stress at these latitudes, their relation with the IWVE and any deviations will be presented in a follow up study. India lies in the Southeast Asian belt where both high temperature and high humidity contribute significantly to heat stress (Im et al., 2017; Hoegh-Guldberg et al., 2018; Wehner et al., 2016) particularly in April, May and June. While IWVE shows an inverse relationship between these variables, their combination could be above the extreme heat wave threshold, as shown by Wehner et al. (2016) for May 2015 in Hyderabad, India. Hence, this region has been chosen for a case study. Record high near surface temperatures of 51 degrees were observed in Phalodi, Rajasthan (27.13°N, 72.36°E) around 19th May 2016 with the heat wave impacting regions pan-India between the months

422 of April and June (Agarwal, 2016; Wu, 2016; Mazdiyasi et al., 2017; Oldenborgh et al.,
 423 2018). The impact of this heat wave on human health was severe as evidenced by a high
 424 number of newspaper reports on heat strokes and human fatalities in this period (Tharoor,
 425 2016). We present a preliminary meteorological analysis of this heat wave to show a pos-
 426 sible relevance of the IWVE during the heat wave period.

427 Sources of tropical heat stress are not as well understood as are for their extra-tropical
 428 counterparts (Perkins, 2015). For example, there are only a handful of studies which ap-
 429 ply the ‘blocking high’ theory of heat waves to these latitudes (Rohini et al., 2016; Rat-
 430 nam et al., 2016; Oueslati et al., 2017). We endeavor to show through this analysis that
 431 the IWVE, identified in the months of April, May and June over the subcontinent, may
 432 also explain a significant portion of the general surface temperature variability, some-
 433 times even during the maximum heat wave period. At this point, following previous mod-
 434 eling studies, it is noted that at around 25° latitude, heat waves associated with extreme
 435 geopotential anomalies may take effect only in a couple of days as compared to almost
 436 a week at higher latitudes (Jiménez-Esteve & Domeisen, 2022; Lau & Nath, 2014). Also
 437 the maximum in T_{max} is found to be co-eval with the maximum in the geopotential height
 438 anomaly (ΔGH) (Jiménez-Esteve & Domeisen, 2022). That is, any lag, of the extreme
 439 T_{max} , more than a couple of days from the development of statistically significant posi-
 440 tive ΔGH may be indicative of other controlling processes. These studies also show that
 441 at 25° latitude, the mid tropospheric ΔGH associated with heat waves could be around
 442 30 m, which can increase to up to 100 m at higher latitudes.

443 Figure 8 shows the May 2016 time series of several meteorological variables over
 444 Phalodi, Rajasthan, India. The variables q_z , pressure velocity at 550 hPa (ω_{550}) and CRF
 445 derived from ERA5, averaged over a $2^\circ \times 2^\circ$ region centered around Phalodi are used to
 446 understand the IWVE. Following studies of extra-tropical heat waves, the geopotential
 447 height anomalies from a climatological mean (1979 to 2020) at 800 hPa and 550 hPa (ΔGH_{800} ,
 448 ΔGH_{550}) are used to represent the role of blocking highs in triggering heat stress. Fig-
 449 ure S7 shows the same variables over other regions in the subcontinent during the same
 450 heat wave. These regions were chosen because they have either a high IWVE, ΔT_{max}
 451 or ΔGH_{550} signal. These figures are summarised in Table 1.

452 The IWVE correlations, either with q_{sfc} or with CWV are moderately high and
 453 statistically significant for all stations during May 2016. The correlation between T_{max}
 454 and ΔGH_{550} is smaller and mostly insignificant at all stations. The highest T_{max} anom-
 455 alies appear around 17th to 19th May, except in Dachehalli in eastern India. The onset
 456 of significant ΔT_{max} usually lags significant ΔGH_{550} by around 4-7 days at all stations
 457 except Dachehalli where the heat wave starts right after a period of heavy precipitation.
 458 Except for Phalodi, no station has a significant ΔGH_{550} on the day of the maxima in
 459 T_{max} . But, for Phalodi, although the period of maximum T_{max} is co-eval with a signif-
 460 icant ΔGH_{550} , the maxima in ΔGH_{550} occurs 10 days earlier. Some stations like Luc-
 461 know and Maharashtra don’t show a persistent multi-day positive ΔGH_{550} preceding
 462 or co-eval with the heat wave. In most cases, the period of significant positive ΔGH_{550}
 463 or ΔT_{max} are not associated with statistically significant downwelling. However, dur-
 464 ing the period of significant ΔT_{max} the CRF is, in general, low in magnitude as com-
 465 pared to the rest of the month or even maximum positive for Gujarat during this period.
 466 The rest of the month is marked by larger negative CRF in general. It is noteworthy
 467 that this period of low CRF does not coincide with the period of significant positive ΔGH_{550} .

468 The above preliminary analysis suggests that statistically significant geopotential
 469 anomalies are not co-eval with the period of highest heat stress at least over the Indian
 470 subcontinent during the severe heat wave of May 2016. In fact, these periods may be sep-
 471 arated from each other by several days. At the same time, every station is found to be
 472 significantly impacted by the IWVE, especially during extremes in surface temperatures.
 473 This observation, although does not preclude the possible role of a blocking high in im-
 474 pacting the heat wave, does provide support for the role of the tropics-wide IWVE in

475 causing regular heat stress throughout the pre-monsoon summer season, every year, con-
 476 tiguously throughout the tropics. This is observed in the Indian subcontinent, which regu-
 477 larly experiences pre-monsoon heat stress (Pai et al., 2013), which could have a dom-
 478 inant component from the continuously varying humidity resulting in extremes in tem-
 479 perature.

480 It should be noted that May 2016 was a severe pre-monsoon heat stress year for
 481 the Indian subcontinent, hence the atmosphere could be impacted by certain synoptic
 482 scale conditions, other than the IWVE, not considered in detail here. Moreover, this anal-
 483 ysis does not consider positive feedback from dry soils which might add another layer
 484 of complexity to local T_{max} evolution (Miralles et al., 2014; Perkins, 2015). Hence, fur-
 485 ther investigation is needed to unravel the relative significance of these processes in caus-
 486 ing heat stress at these latitudes.

487 4 Conclusion

488 This study shows the existence of regions at the borders of the inter-tropical con-
 489 vergence zone where the atmospheric dynamics seems to specifically support the occur-
 490 rence of clouds with a negative radiative impact on daily maximum surface temperatures
 491 T_{max} in the current climate. Additionally, other clouds, known to have a positive radi-
 492 ative forcing, have tapering down occurrences in these same regions. Both, cloud areal
 493 coverage and surface cloud radiative forcing (CRF) have a positive correlation with lower
 494 tropospheric water vapor q_{sfc} in these regions. As a result, surface temperatures are strongly
 495 negatively correlated with water vapor. These regions, where water vapor has an appar-
 496 ent negative impact on surface warming, have been referred to as the regions of ‘indi-
 497 rect water vapor effect’ or IWVE.

498 Such a negative forcing from water vapor is not observed in any other extensive
 499 region outside this belt around the ITCZ which coincides with the horizontal branch of
 500 latitudinal overturning circulations, like the Hadley cell. The negative daily CRF may
 501 not be unique to the IWVE regions, however the strong positive correlation of q_{sfc} with
 502 CRF and negative correlation in turn with T_{max} is uniquely observed in the IWVE re-
 503 gions only. We suspect that in other regions, processes other than IWVE which impact
 504 T_{max} through q_{sfc} , may become important, but were not the focus of this study. The
 505 mechanism which lets water vapor to be dominantly, positively correlated with cloud oc-
 506 currence of low and mid-level clouds in these regions, remains to be explored.

507 The IWVE regions occur on both sides of the ITCZ in all seasons and follow its
 508 annual north-to-south march. This means, periods where IWVE is active, both precede
 509 and succeed the monsoon in a tropical region. Although not addressed in detail in this
 510 study, this can have consequences for pre-monsoonal heat stress in regions like the Sa-
 511 hel in Africa, the Cerrado in South America and the Indian subcontinent where the IWVE
 512 is found active in this season which is already fraught with positive surface temperature
 513 feedback from receding soil moisture (Perkins, 2015; Pai et al., 2013; Miralles et al., 2014;
 514 Oueslati et al., 2017). The post monsoon occurrence of the IWVE may not have a sub-
 515 stantial health impact because of the general lower surface temperatures due to wet soils
 516 and a receding sun after the monsoons. An in-depth analysis of the role of IWVE ver-
 517 sus other mechanisms of pre-monsoon heat stress in these regions is required.

518 Lastly, The IWVE seems to have a connection with the horizontal branch of lat-
 519 itudinal overturning circulations in the tropics, in the way the latter affects cloud dis-
 520 tribution. With evidence that tropical overturning circulations, like the Hadley cell, are
 521 set to change in a warmer climate (Lu et al., 2007), the IWVE and its associated impact
 522 on surface temperatures may evolve as well. An understanding of how this sensitive bal-
 523 ance would be modified under global climate change could be important, especially be-

524 cause in the IWVE regions, this effect may be substantially associated with pre-monsoonal
525 heat stress.

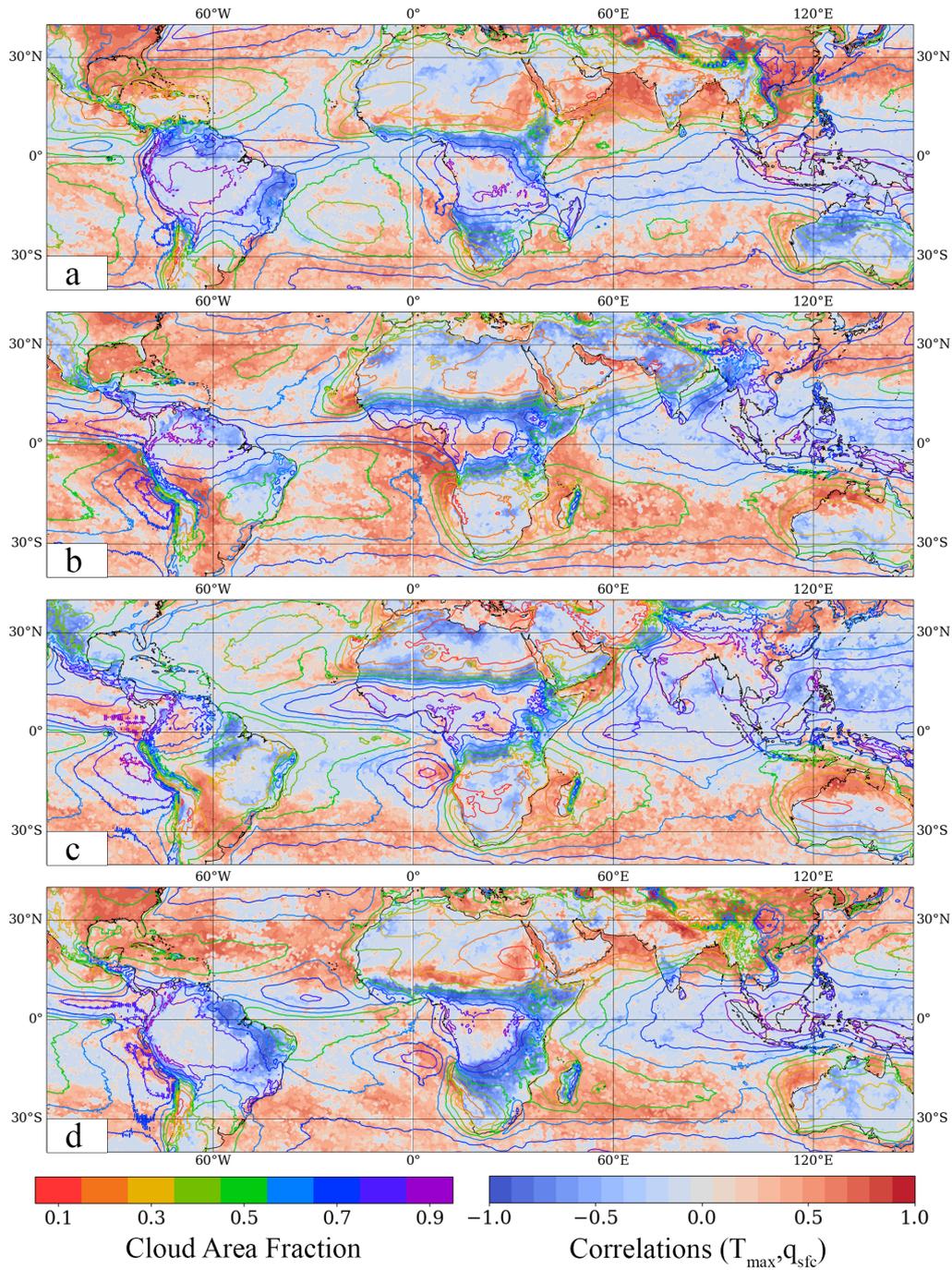


Figure 1. The tropical indirect water vapor negative radiative effect and its co-occurrence with transition regions in cloud fraction. Colored Shading - Mode of monthly daily scale correlations between daily T_{max} and q_{sfc} in (a) February, (b) May, (c) August, and (d) November. The mode was calculated from the population of all monthly correlations between 2001-2020. See methods for details. Colored line Contours – Average total cloud area fraction for the corresponding months between 2001 and 2020. All data from ERA5.

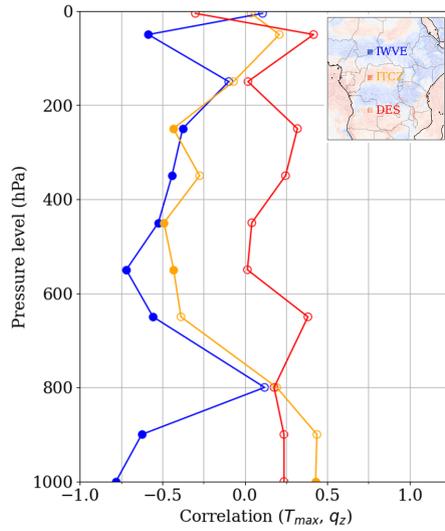


Figure 2. IWVE regions show negative correlations between T_{max} and upper tropospheric specific humidity. Correlation coefficients between daily T_{max} and daily average specific humidity at various altitudes, q_z , in May 2016 at selected IWVE, ITCZ and DES regions in Africa (inset). Data from ERA5. Correlations significant at the 95% confidence level are shown with filled circles.

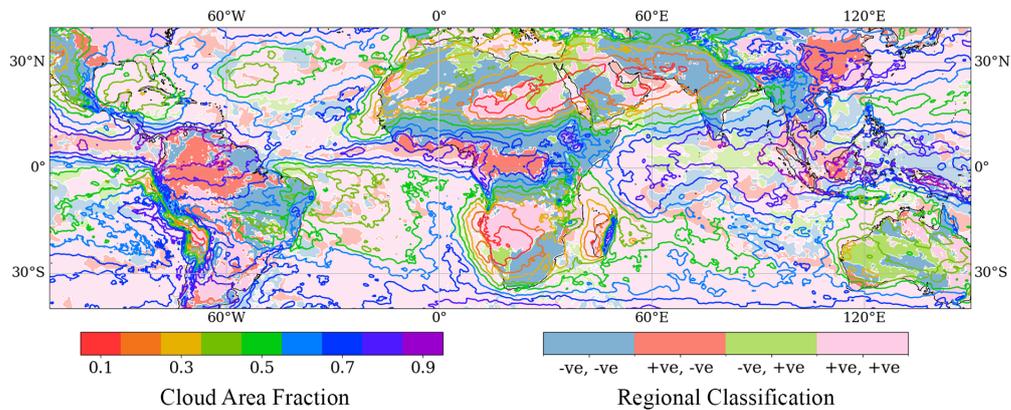


Figure 3. Dominant IWVE regions in the tropics. Classification of various regions into four types based on whether the sign of the grid scale correlations between daily T_{max} and q_{sfc} is the same as that between T_{max} and CWV . Area 1 - both negative, Area 2 - former positive, latter negative, Area 3 - negative, positive, Area 4 - both positive. Data obtained from ERA5 for May 2016 (compare this with Figure S1b). The map has an overlay of average cloud cover for May 2016. Ocean grid points have been de-saturated.

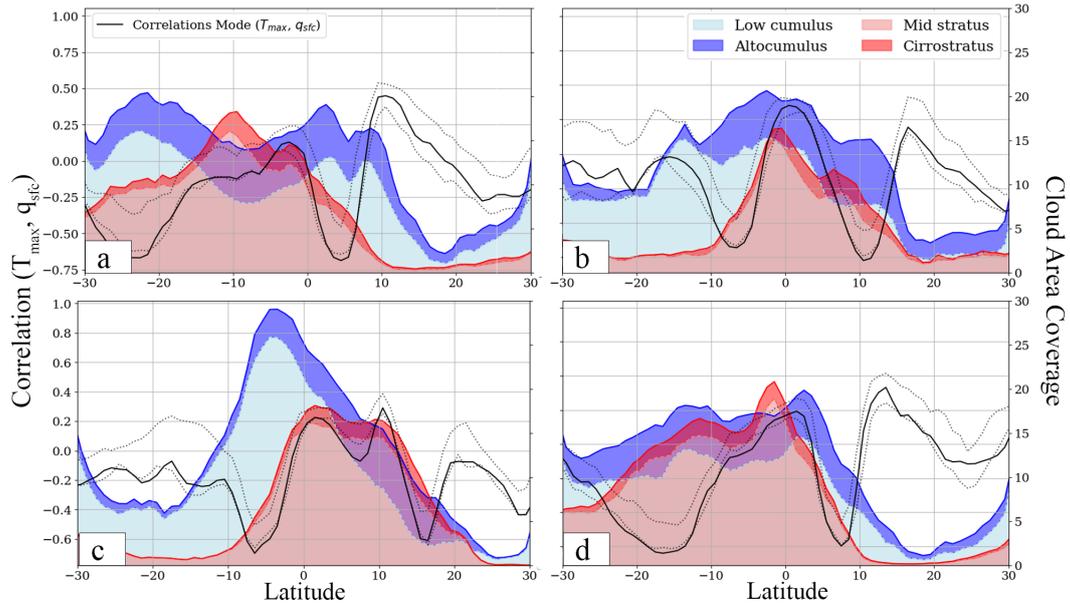


Figure 4. Dominant presence of mid level clouds in IWVE regions. Daytime cloud fraction over Africa, zonally averaged between 30°N, 17.5°E and 30°S, 30°E for (a) February, (b) May, (c) August, and (d) November, 2001 to 2020. Cloud types shown - Low cumulus including shallow cumulus (low, thin) and stratocumulus (low, medium thick); altocumulus (mid, thin); mid stratus including altostratus (mid, medium thickness) and nimbostratus (mid, thick); cirrostratus (high, medium thick). Data from CERES CldTypHist 1°×1° monthly average day time values. Cumulus and stratus clouds with the largest % area coverage have been shown, while other cloud species such as deep convection, nimbostratus and cirrus clouds have been omitted from this analysis due to their comparatively low area coverage (see Fig. S5). This is done assuming that clouds with small % area cover do not have dominant radiative effect at the surface.

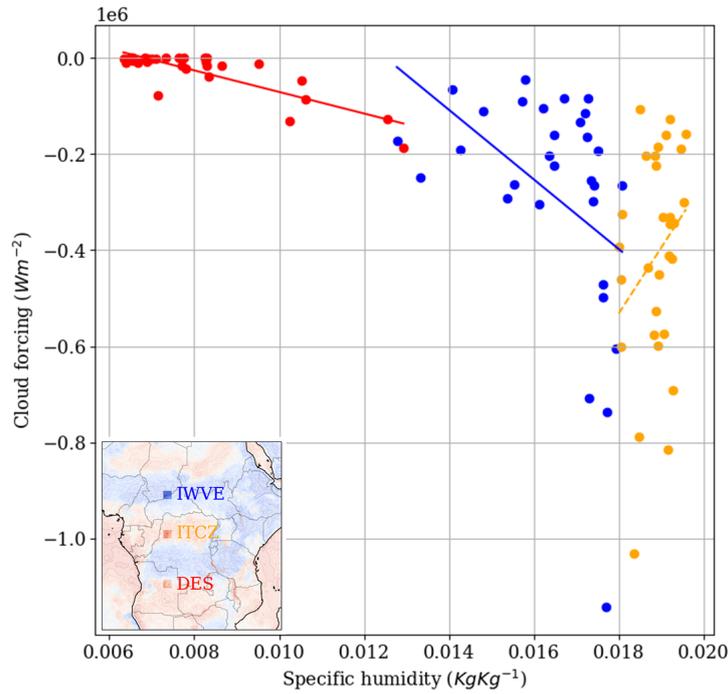


Figure 5. Surface cloud radiative forcing in IWVE regions is sensitive to near surface atmospheric humidity. Daytime averaged CRF versus q_{sfc} for a $1^\circ \times 1^\circ$ IWVE, ITCZ and a DES region in Africa (inset), for all days in May 2016. All data from ERA5.

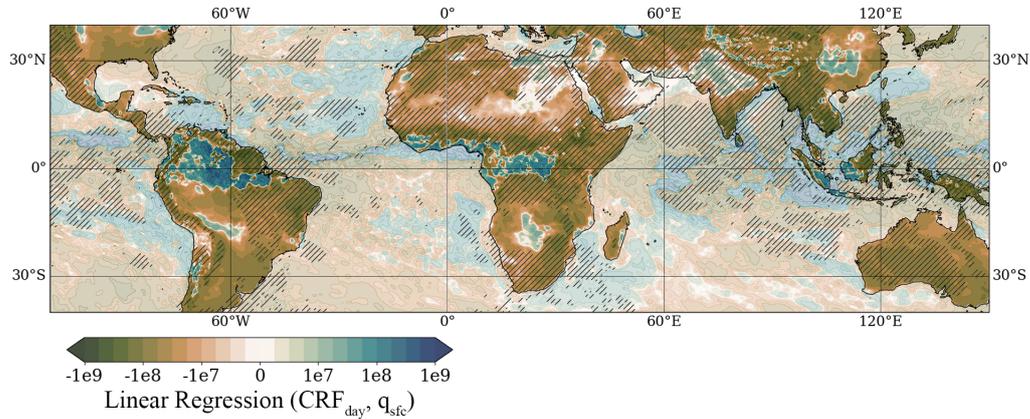


Figure 6. IWVE across tropics show a negative relation of surface humidity with cloud radiative forcing. Coefficients of linear regression of daytime CRF on daily averaged q_{sfc} . Regression performed with ERA5 grid scale, daily data for May 2016. Most values above $-7e7$ are significant at the 95% confidence level. Most positive regression coefficients are insignificant. Statistical significance is not shown to improve figure clarity. Hatching shows all regions with negative correlations between T_{max} and q_{sfc} . Data obtained from ERA5.

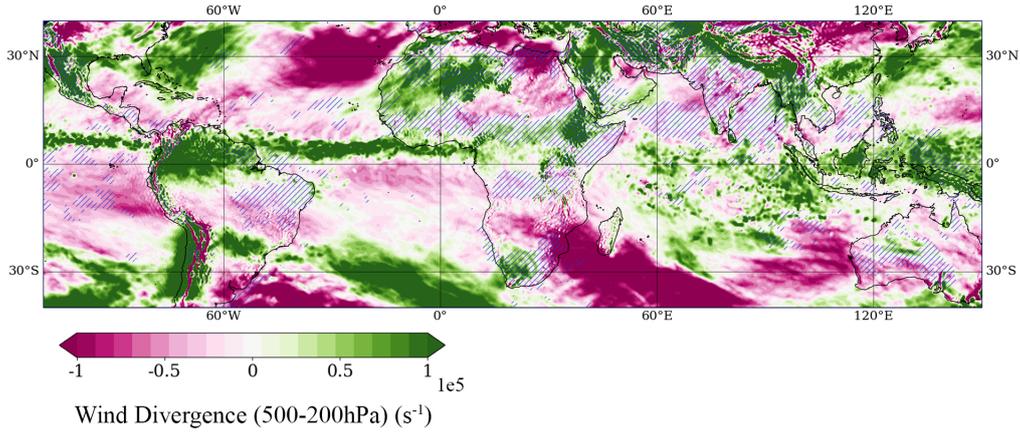


Figure 7. IWVE associated with the horizontal branch of continental scale overturning circulations. Shading shows the divergence of horizontal winds, column integrated between 500 hPa and 200 hPa averaged over May 2016. Hatching shows IWVE regions with T_{max} , q_{sfc} correlation coefficients between -1 and -0.5. Data obtained from ERA5.

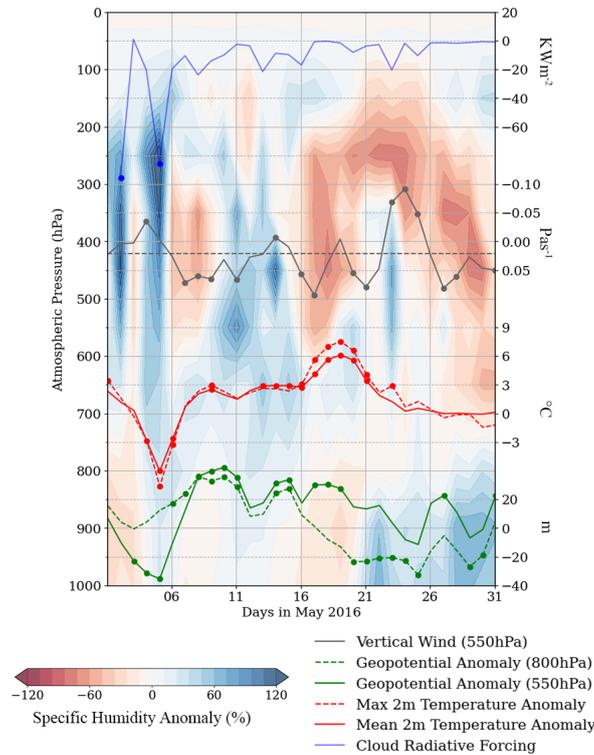


Figure 8. Severe tropical heat stress could be related to IWVE. Daily scale time series of T_{max} , T_{avg} (daily mean), GH_{800} and GH_{550} anomalies from the climatological mean corresponding to May 1979 to 2020. Figure also shows the percentage temporal anomalies in q_z from the monthly mean of May 2016, the daytime CRF and daily 550 hPa vertical wind ω_{550} . The dashed grey line represents the climatological average vertical wind in May. Significant anomalies are marked. Significant differences of CRF and ω_{550} from the monthly and climatological mean respectively are also marked. All variables derived from a $2^\circ \times 2^\circ$ area over Phalodi, north-western India in May 2016. Data obtained from ERA5.

Table 1. Importance of IWVE during the heat wave of May 2016 over the Indian subcontinent. Meteorological conditions during the May 2016 heat wave recorded in India. Statistics calculated using a month long daily scale time series of meteorological variables around centers of heat stress, with the same methodology as used in Figure 8. Meteorological conditions for two types of days are analyzed - day of the monthly maximum T_{max} and the day of the monthly maximum geopotential anomaly at 550 hPa, ΔGH_{550} . For the heat wave day values are reported for the daily maximum and averaged surface temperature anomalies from the climatological (1979 to 2020) monthly mean ΔT_{max} and ΔT_{avg} , daily scale IWVE correlations $\text{corr}(T_{max}, q_{sfc})$ and $\text{corr}(T_{max}, CWV)$, daily scale area averaged vertical wind at 550 hPa ω_{550} , daytime averaged surface CRF , anomaly in the geopotential heights at the 800 hPa and 550 hPa levels during the heat wave period (ΔGH_{800} , ΔGH_{550}), and correlations $\text{corr}(T_{max}, GH_{800})$ and $\text{corr}(T_{max}, GH_{550})$. Some of these values are also reported for the day of maximum ΔGH_{550} . Statistics significant at the 95% confidence level are bold. All data from ERA5. All values are averaged over a $2^\circ \times 2^\circ$ area around the region under study.

	Phalodi	Datia	Lucknow	Dachepalli	Radhanpur	Patoda
latitude ($^\circ\text{N}$)	27.13	25.67	26.87	16.5	23.75	18.75
longitude ($^\circ\text{E}$)	72.36	78.45	80.94	79.75	71.75	75.5
$\text{avg}(T_{avg, May})$	34.96	34.05	33.78	33.2	33.7	32.38
Day of max(T_{max})	19	18	17	24	18	17
ΔT_{max} ($^\circ\text{C}$)	7.51	3.97	4.26	4.16	6.14	3.79
ΔT_{avg} ($^\circ\text{C}$)	6.1	3.83	3.54	4.05	3.99	3.8
$\text{corr}(T_{max}, q_{sfc})$	-0.4	-0.37	-0.47	-0.81	-0.37	-0.45
$\text{corr}(T_{max}, CWV)$	-0.62	-0.41	-0.57	-0.67	-0.69	-0.24
ω_{550} (Pa sec^{-1})	0.0	0.0	0.01	0.03	0.03	0.03
CRF (kW m^{-2})	-1.51	0	-19.1	-33.39	0.5	-18.97
ΔGH_{800} (m)	-13.16	-14.83	-11.04	3.86	-10.91	-20.2
ΔGH_{550} (m)	27.6	14.37	6.27	7.62	19.74	-2.96
$\text{corr}(T_{max}, GH_{800})$	-0.2	-0.39	-0.17	0.33	0	-0.13
$\text{corr}(T_{max}, GH_{550})$	0.38	0.17	0.32	0.39	0.4	0.31
Day of max(ΔGH_{550})	10	14	14	14	14	14
ΔT_{max} ($^\circ\text{C}$)	2.27	2.73	3.11	1.01	2.43	2.38
ΔGH_{550} (m)	42.33	35.53	35.22	29.9	32.01	34.9
CRF (kW m^{-2})	-9.98	-31.9	-30.56	-22.33	0.28	-76.91

5 Open Research

[H] The meteorological data used for the heat stress analysis in the study is available freely on the Copernicus Climate Data Store via 10.24381/cds.143582cf, <https://cds.climate.copernicus.eu/#!/search?text=ERA5&type=dataset> (Hersbach et al., 2020, 2023). Supporting meteorological data from the AIRS L3 standard data products V7 is obtained from the data collections at NASA GES DISC via 10.5067/UO3Q64CTTS1U, <https://disc.gsfc.nasa.gov/datasets?page=1&source=AQUA%20AIRS&keywords=airs%20version%207> (AIRS, 2019). CERES cloud type and surface radiation flux data are obtained from the SYN1deg level 3 products from the NASA CERES data repository via (doi - 10.5067/TERRA+AQUA/CERES/CLDTPHIS.L3.004, doi - 10.5067/Terra+Aqua/CERES/SYN1degDay.L3.004A respectively) <https://ceres.larc.nasa.gov/data/> (Doelling et al., 2013, 2016). All datasets used are available freely for public use.

All original Python (> 3.9) code will be archived if the manuscript is accepted and can be obtained from the repository or upon request from Karan at karan42mahajan@gmail.com. All data used from the above sources can also be separately obtained from Karan.

Acknowledgments

The authors acknowledge financial support from the National Institute of Science Education and Research, Bhubaneswar, India and the Ministry of Education, Government of India, grant number MoE-STARS1/766.

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