# An indirect, negative radiative effect of water vapor in the tropics and its implications for regional surface 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.

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

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Key Points:

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9	An indirect, water vapor related surface cooling is observed at the daily scal	le, 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 sea	sonal

cycle, possibly consequential for pre-monsoon heat stress

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#### <sup>35</sup> Plain Language Summary

The greenhouse radiative effect of water vapor on surface temperatures is widely 36 known. However, can water vapor cause an 'apparent' radiative cooling effect under cer-37 tain conditions? Here, we show that in certain extensive tropical regions, bordering the 38 ITCZ, the lower tropospheric water vapor has a substantial relation with cloud occur-39 rence, which subsequently renders an indirect surface cooling effect in high humidity con-40 ditions. These clouds are majorly shallow and mid level cumuli, which are known to have 41 a negative surface radiative forcing. What causes the positive relation between lower tro-42 pospheric humidity and these clouds, is not understood. However, that these regions bor-43 der the ITCZ and hence are co-located with the horizontal branch of the Hadley cell, sug-44 gests low amounts of convective triggering and hence high level clouds. The potentially 45 reduced warming from the high level clouds may give the lower level clouds an upper hand 46 in the net radiative impact. Lastly, these regions precede the ITCZ in time as they move 47 north-south with it. This may be consequential for pre-monsoonal heat stress in trop-48 ical regions. 49

#### 50 1 Introduction

Water vapor is frequently dubbed as the most important greenhouse gas which plays 51 a dominant role in climate change through the water vapor feedback (Soden et al., 2002, 52 2005; Forster et al., 2021; Dessler et al., 2013). This control is delivered through the dom-53 inant impact of the increase in the upper tropospheric humidity on trapping the outgo-54 ing longwave radiation (OLR) and recycling it within the climate system (Soden et al., 55 2005; Allan et al., 1999; Dessler et al., 2013). This trapping of energy warms up all com-56 ponents of the climate system - the atmosphere, land and oceans. However, on hourly 57 to daily time scales, it is the near surface humidity (or column integrated water vapor, 58 as they are strongly correlated) which dominates the greenhouse effect of water vapor 59 as manifested at the land surface (Shakespeare & Roderick, 2021; Flerchinger et al., 2009; 60 M. Li et al., 2018; Brutsaert, 1975; Carmona et al., 2014). Idealized modeling studies 61 of clear sky surface downwelling longwave radiation (DLR) are able to explain about 97% 62 of the temporal variability in DLR with near surface humidity  $(q_{sfc})$  and surface tem-63 peratures  $(T_{sfc})$  (Shakespeare & Roderick, 2021). High positive correlations have been 64

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

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

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

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

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 124 with local cloud cover, on daily scale surface heat stress in tropical regions during the 125 peak, pre-monsoon, summer season. The general correlations between the humidity and 126 surface temperature variables are presented first, subsequently relating the observation 127 to the local cloud cover and variability. Next, the relationship of these correlations with 128 the larger scale atmospheric dynamics is presented. Lastly, a short analysis is presented 129 to propose a possible role of the near surface humidity in causing observed heat stress 130 through the processes identified. 131

#### 132 2 Methods

This study focuses on processes that impact the daily scale variability of regional 133 surface temperatures  $(T_{sfc})$ . As such, most analysis uses daily scale data for a selected 134 month. This choice of time scales reflects our interest in the 'instantaneous' response of 135  $T_{sfc}$  to the radiative forcing by the variability in water vapor and clouds. Since the lat-136 ter process takes of the order of a few hours to manifest, the day-to-day co-variability 137 in the variables is analyzed, if not the co-variability at sub-daily time scales. The atmo-138 spheric water vapor variability at longer time scales might itself be caused by inter-seasonal 139 and inter-annual climate variability and climate change, but that is not the focus of this 140 study. However, our analysis with inter-annual, monthly averaged variables results in sim-141 ilar inferences as the analysis with daily scale variables, providing support for the gen-142 eralizability of the processes identified. Results are majorly reported from the pre-monsoon 143 summer season of the northern hemisphere, but the inferences remain unchanged for other 144 seasons. The month of May 2016 is chosen for some specific analysis that is limited by 145 the size of the data set involved. 2016 was a post El Niño year transitioning to a mild 146 La Niña, infested by high heat stress in the Indian subcontinent (Oldenborgh et al., 2018). 147

Surface heat stress is quantified with daily scale maximum surface temperatures 148  $(T_{max})$  from reanalysis and satellite datasets.  $T_{max}$  appears frequently and takes an im-149 portant role in the formal estimation of heat stress (IMD, 2002; Perkins, 2015). How-150 ever, in this study, heat stress is used to refer to periods of statistically significant de-151 partures of  $T_{max}$  from the climatological or monthly mean and should not be confused 152 with the various heat wave indices used by meteorological departments to declare pe-153 riods of heat stress. Most inferences reported here with  $T_{max}$  remain unchanged for an 154 analysis with daily averaged surface temperatures  $(T_{avg})$  and daily maximum 2 m air tem-155 perature  $(T_{2m,max})$ . 156

The daily averaged near surface specific humidity  $(q_{sfc})$ , specific humidity at var-157 ious altitudinal levels  $(q_z)$  and total column water vapor (CWV) are used to capture the 158 radiative effect of atmospheric water vapor on surface temperatures. Previous studies 159 have utilized these variables to show that at daily to monthly timescales, in most global 160 land areas these variables are non-linearly but positively correlated with the surface down-161 welling longwave radiation (DLR) or surface temperatures (Gaffen et al., 1992; Ross et 162 al., 2002). Although the clear sky regression coefficients between  $q_{sfc}$  and DLR are space 163 dependent (Allan et al., 1999), there is no previous evidence of them being negative. Hence, 164  $q_{sfc}$  is used to represent the greenhouse impact of atmospheric water vapor on surface 165 temperatures (Shakespeare & Roderick, 2021). 166

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

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

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

Most analysis in this work is performed using statistics like linear correlation coefficients, linear regressions coefficients and anomalies from spatio-temporal averages for daily scale data analyzed over a month. Statistical significance is obtained with a twotailed t-test for correlation coefficients, linear regressions and differences at the 95% confidence level in all cases.

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

Figures 1 and S1 show the multiyear mode of monthly correlations of daily scale  $T_{max}$  (or  $T_{2m,max}$ ,  $T_{avg}$ ) and  $q_{sfc}$  (or CWV). Multiyear mode is calculated using gridscale monthly correlations for 20 years, from four neighboring grid points, regardless of the statistical significance of the correlations in individual years. However, the correlations of interest (IWVE correlations defined in 'Results and Discussion') are robustly significant across all years. Fig. S1b is shown as an example and will be discussed in detail in the next section.

#### <sup>209</sup> **3** Results and Discussion

The objective of this study is to identify regions where water vapor impacts surface heat stress through processes over and above its greenhouse warming impact. In other words, we endeavor to investigate regions where the direct greenhouse effect of water vapor may be masked by the radiative impacts of the other climatically important processes mediated by water vapor, like cloud formation. To identify such regions, a basic correlation analysis between daily scale surface temperature and indicators of atmospheric humidity is performed. 217 218

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

At the daily scale,  $T_{max}$  of a majority of regions around the globe are strongly pos-219 itively correlated with  $q_{sfc}$  (Fig. 1), and CWV (Fig. S1 a) in ERA5 and also in AIRS 220 (Fig. S2a). The values of these correlation coefficients for May 2016, are displayed in 221 Fig. S1 b to show that the spatial patterns of the multiyear mode of these correlations 222 (Figs. 1, S1) compare well with the values for individual years. These widespread pos-223 itive correlations could be attributed to the greenhouse effect of water vapor. The max-224 225 imum positive correlations larger than  $\approx 0.7$  are obtained at high latitudes and moderate values over the tropical oceans, a feature of the surface temperature-humidity rela-226 tionship reported before (Ross et al., 2002). The inhibition of these positive correlations 227 along the inter-tropical convergence zone (ITCZ) over land could be attributable to the 228 dominant presence of high level clouds in these regions, which may have a complex net 229 *CRF*. The analysis of these regions is out of the scope of this study. 230

The same figures, however, also show several tropical and sub-tropical land regions 231 with high, statistically significant, negative correlations between daily  $T_{max}$  and  $q_{sfc}$  which 232 occur across seasons and the two data sets used. Such signal has previously been observed 233 over western USA using radiosonde data (Ross et al., 2002). In our study the negative 234 correlation coefficients range between -0.36 to -0.97 for data averaged for May between 235 2001 and 2020 (Fig. 1). The largest negative correlations are found predominantly over 236 tropical Africa and tropical South America throughout the year and over the Indian sub-237 continent, southeast Asia and northern Australia in some seasons. Similar patterns ap-238 pear also with  $T_{avg}$  and  $T_{2m,max}$  (Figs. S1 c and S1 d) and with inter-annual correlations 239 using monthly averaged data (Fig. S2b), indicating the generalizability of this relation-240 ship. This apparent water vapor related cooling is observed over land areas on the flank 241 regions of the ITCZ and follows its seasonal cycle (Fig. 1). Hence, for example over the 242 Indian subcontinent, it comes into prominence in the dry, pre-monsoon summer season 243 (Fig. S3). It should be noted however that these negative correlations might maximize 244 in a region's peak monsoon period. This is probably due to an atmospheric moistening 245 and a sharp evaporative surface cooling due to heavy precipitation - for example in June 246 over India when the monsoons arrive over land (Fig. S3). However, the cause of the wa-247 ter vapor related cooling in the pre-monsoon dry period is not obvious. 248

This negative correlation of near surface water vapor with surface temperatures, 249 suggesting an apparent negative radiative forcing from water vapor is termed here as the 250 'Indirect Water Vapor Effect' or IWVE. This apparent indirect forcing of water vapor 251 is investigated next. In the following, IWVE correlations will mean a correlation between 252 daily  $q_{sfc}$  and  $T_{max}$ , unless otherwise specified. Furthermore, the positive IWVE cor-253 relation regions over the deep convective regions near the equator are termed as the ITCZ 254 and the positive IWVE correlation regions polewards of the IWVE regions are termed 255 as the desert or DES regions. 256

Figure 2 shows the correlations between specific humidity at an altitude  $(q_z)$  and 257 daily  $T_{max}$  over three regions of the African subcontinent - an IWVE, an ITCZ and a 258 DES region in May 2016. These regions are carefully chosen to lie at the same longitude 259 (19°E) and away from topography. Over the IWVE region, the correlation between  $T_{max}$ 260 and  $q_z$  is always significantly negative throughout the troposphere. Similarly, over the 261 DES region the above correlation is significantly positive throughout. However, over the 262 ITCZ region the correlation is positive close to the surface but changes to negative in 263 the upper half of the troposphere. The altitudinal dependence of the correlations over 264 265 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. 266 Only the IWVE regions, where  $q_{sfc}$  dominates the overall correlation between CWV and 267  $T_{max}$ , are studied here. This is a reasonable choice because most IWVE regions defined 268

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

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# 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 273 indicated time periods from ERA5. The IWVE co-occurs with regions of transition cloud 274 cover, tropics-wide, i.e. between regions of maximum and minimum cloud cover where 275 its gradient is substantial. For example, during May over central Africa (Fig. 1b), the 276 average cloud cover over the IWVE regions is around 0.51 (standard deviation 0.17), as 277 compared to around 0.71 (0.18) over the ITCZ and 0.24 (0.18) over the DES regions. Hence, 278 the total cloud cover in the IWVE regions is intermediate between regions of predom-279 inant convection (ITCZ) and regions of predominant downwelling (over subtropical deserts 280 like the Sahara, Kalahari, the Arabian desert etc). Additionally, the fact that the IWVE 281 regions tightly border the equatorial regions of deep convection everywhere, and follow 282 the north-south march of the ITCZ, suggests that the IWVE correlations could be me-283 diated by the radiative effects of some specific clouds that are dominant at the borders 284 of the ITCZ. 285

Next, a possible pathway, mediated by clouds, is analyzed, which may lead to the 286 apparent negative radiative forcing from water vapor in the IWVE regions. These re-287 gions receive much lower precipitation than the ITCZ (Fig. S4) which suggests that the 288 IWVE regions may harbor clouds, or a cloud distribution, that is distinctly different from 289 those in the ITCZ, resulting in a distinctly different CRF. Clouds show a spectrum of 290 surface and top of atmosphere CRF depending upon their cloud top height and opti-291 cal depth (Chen et al., 2000; Hartmann et al., 1992). A radiative transfer calculation is 292 often necessary to obtain the net radiative effect of a certain vertical distribution of clouds. 293 However, in this study we have taken evidential support from previous research to tie 294 the observed cloud distribution over the IWVE regions with the negative, apparently cloud-295 mediated radiative impact of  $q_{sfc}$ , both of which are discussed below. A comprehensive 296 numerical analysis of these relationships is left for a follow up study. 297

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#### 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 zon-299 ally averaged over a region in Africa  $(30^{\circ}N, 17.5^{\circ}E \text{ and } 30^{\circ}S, 30^{\circ}E)$ , obtained from the 300 CERES CldTypHist product for May 2001 to 2020. The corresponding May average cloud 301 maps for Africa are provided in Figure S5. There is a clear distinction in the distribu-302 tion 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 mod-304 erately thick clouds, like shallow cumuli, altocumuli and stratocumuli. While the ITCZ, 305 apart from the low and mid level clouds, also have a higher percentage of optically thin 306 to thick, mid and high clouds, relative to the IWVE regions. This feature is present in 307 all seasons except when the IWVE correlations are large during the monsoon, apparently 308 due to the presence of strong evaporative surface cooling (August in Northern Hemisphere 309 in Figure 4c), in which case the low and altocumulus clouds become infrequent over the 310 IWVE regions. Specifically, the non-monsoonal, IWVE regions show a dominant pres-311 ence of mid level, low thickness altocumulus clouds, which follow the poleward bound-312 aries of the IWVE regions and are present abundantly and uniformly between these lat-313 itudes (Figure S5d). 314

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 (opti-319 cal depth around 1.03) mid level clouds (3 to 8 km) with a possibly negative radiative 320 forcing over land has been previously reported in the pre-monsoon months over west-321 ern Africa in the Nigerian region (Bouniol et al., 2012; Bourgeois et al., 2016; Stein et 322 al., 2011), in Darwin (Riihimaki et al., 2012) and in the pre-monsoon Indian subconti-323 nent (Parker et al., 2016). These mid level, thin, altocumulus clouds with a negative ra-324 diative forcing in the tropical regions (Bourgeois et al., 2016), are associated with mid 325 level detrainment from deep convective clouds near the zero degree isotherm (Johnson 326 et al., 1996, 1999; Iwasa et al., 2012). It is suggested that the latent heat of condensa-327 tion to ice causes the deep convective tower to lose buoyancy and form these clouds, which 328 can get horizontally advected to far away regions. As such, the occurrence and radia-329 tive impact of such mid level clouds is likely to be important in the flank regions of the 330 equatorial deep convective belt, which is where the IWVE regions occur. 331

Hence, the dominant shallow- and alto- cumulus clouds in the IWVE regions could 332 be responsible for the observed surface cooling. However, the question - how is  $q_{sfc}$  in 333 the IWVE regions linked to surface cooling caused through these clouds? - is still open. 334 This is an important question, because the expected positive, greenhouse radiative forc-335 ing of water vapor seems to be negated by a mechanism that supports clouds with a neg-336 ative radiative impact. This mechanism should cause the aforementioned cooling clouds 337 to occur along with high  $q_{sfc}$  in the IWVE regions. Given the strength and robustness 338 of the IWVE correlations over several time and spatial scales, this mechanism could have 339 great significance for the heat stress dynamics in the IWVE regions. Although the ex-340 act mechanism is not investigated here, we show evidence of such a negative relation-341 ship between CRF and  $q_{sfc}$  that exists dominantly over the IWVE regions, rendering 342 the 'effective' negative radiative forcing from water vapor. 343

Over the IWVE regions, we find a robust, significant and large negative (positive) 344 regression between  $q_{sfc}$  and daytime CRF (cloud fractional coverage, figure not shown), 345 but insignificant regression between these quantities in the DES regions and positive, in-346 significant values in some areas of the ITCZ (Fig. 5). This feature of the IWVE, ITCZ 347 and DES regions is observed pan-tropics (Figures 6 and S6) throughout the year (fig-348 ures not shown). However, we do find the ITCZ regions to have a significant negative 349 correlation between upper tropospheric water vapor and CRF (figure not shown) indi-350 cating a dependence of cloud formation on local environmental humidity, as is presum-351 ably also the case over the IWVE regions. In the IWVE regions, we however find this 352 strong dependence of CRF on the water vapor present throughout the column (figure 353 not shown). In other words, in the IWVE regions, water vapor in the lower and higher 354 tropospheric levels seems to vary and impact cloud cover in tandem. The mechanism be-355 hind this apparent correlation is not understood, however, the correlation enables lower 356 tropospheric water content to be a good indicator of CRF in these regions. 357

Lastly, the IWVE regions also show a large day-to-day variability in both cloud cover 358 and CRF, which is not present in the DES and ITCZ regions (Fig. 5). Over DES,  $q_{sfc}$ 359 is too low to be associated with substantial cloud occurrence, whereas, over the ITCZ 360 it is too high to limit or control the same. While we have not established any cause-and-361 effect relationships between  $q_{sfc}$  and cloud cover in the three regions, the general atmo-362 spheric conditions do seem to be less conducive to or supportive of clouds over the DES 363 regions and vice versa over the ITCZ. Some of this possibly could be attributed to the 364 stable conditions that form over the DES regions due to Hadley circulation and vice versa 365 over the ITCZ. The IWVE regions however show a large variability in both  $q_{sfc}$  and CRF. 366 suggesting atmospheric conditions variable enough to support different types and amounts 367 of clouds. This feature is observed in most IWVE regions throughout the tropics (fig-368 ure not shown). The effect of tropical latitudinal overturning circulation on the IWVE 369 regions is analyzed next. 370

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## 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 380 the year due to maximum solar insolation, which results in the high cloud cover and pre-381 cipitation associated with these regions discussed earlier. The divergence of the horizon-382 tal wind, column integrated between 500 hPa and 200 hPa in the tropics, shows positive 383 values (divergence) over the ITCZ regions (Fig. 7). Hence, the regions where  $q_{sfc}$  is pos-384 itively (not) correlated with  $T_{max}$  (CRF) are also the regions of deep convective forc-385 ing. The high cloud cover of all types in these regions supports this inference (Fig. 4). 386 The opposite is true for DES regions which show predominant horizontal convergence, 387 downwelling and clear skies (Fig. 7). 388

The IWVE regions, on the other hand, are marked by the smallest horizontal di-389 vergence and vertical motion as depicted in Figure 7. This is seen by the overlap between 390 the IWVE regions and the contours of near zero divergence of the horizontal winds in-391 tegrated over the upper troposphere. This feature is present over all IWVE regions in 392 the tropics. This absence of a strong vertical forcing could be related with the dominance 393 of shallow cumulus and altocumulus clouds and the lower areal coverage of deep convec-394 tive and cirrus clouds in the IWVE regions (Figure 4). It can only be speculated now 395 and verified in a numerical study that vertical forcing is conducive to a certain type of 396 clouds, the absence of which may be related with the IWVE. 397

398

#### 3.5 Why could the water vapor indirect effect be important?

The co-location of the IWVE regions with transition cloud cover and negligible ver-399 tical motion (or convective forcing) indicate that these regions occur in the horizontal 400 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 402 find a strict latitudinal dependence of the IWVE regions over various continents in a given 403 season. In fact, continental scale overturning circulations, like the one found over the north-404 ern Sahara in May 2016 (Figure 7), can also engender IWVE regions co-located with their 405 horizontal branches. However, Figure 1 demonstrates clearly that the IWVE is not present 406 in regions between Libya and Mali over a climatological average and may not be impor-407 tant for the regional heat stress climatology. However, in regions affected by a climato-408 logical latitudinal overturning circulation, like the Hadley cell, the IWVE could be de-409 cisive in determining pre-monsoonal surface heat stress. 410

To this effect, we provide an example of the surface heat stress during the severe, 411 pan-India heat wave of May 2016. An exhaustive analysis of the general heat stress at 412 these latitudes, their relation with the IWVE and any deviations will be presented in a 413 follow up study. India lies in the Southeast Asian belt where both high temperature and 414 high humidity contribute significantly to heat stress (Im et al., 2017; Hoegh-Guldberg 415 et al., 2018; Wehner et al., 2016) particularly in April, May and June. While IWVE shows 416 417 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 Hyder-418 abad, India. Hence, this region has been chosen for a case study. Record high near sur-419 face temperatures of 51 degrees were observed in Phalodi, Rajasthan (27.13°N, 72.36°E) 420 around  $19^{th}$  May 2016 with the heat wave impacting regions pan-India between the months 421

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

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

Figure 8 shows the May 2016 time series of several meteorological variables over 443 Phalodi, Rajasthan, India. The variables  $q_z$ , pressure velocity at 550 hPa ( $\omega_{550}$ ) and CRF444 derived from ERA5, averaged over a  $2^{\circ} \times 2^{\circ}$  region centered around Phalodi are used to 445 understand the IWVE. Following studies of extra-tropical heat waves, the geopotential 446 height anomalies from a climatological mean (1979 to 2020) at 800 hPa and 550 hPa ( $\Delta GH_{800}$ ) 447  $\Delta GH_{550}$ ) are used to represent the role of blocking highs in triggering heat stress. Fig-448 ure S7 shows the same variables over other regions in the subcontinent during the same 449 heat wave. These regions were chosen because they have either a high IWVE,  $\Delta T_{max}$ 450 or  $\Delta GH_{550}$  signal. These figures are summarised in Table 1. 451

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

The above preliminary analysis suggests that statistically significant geopotential anomalies are not co-eval with the period of highest heat stress at least over the Indian subcontinent during the severe heat wave of May 2016. In fact, these periods may be separated from each other by several days. At the same time, every station is found to be significantly impacted by the IWVE, especially during extremes in surface temperatures. This observation, although does not preclude the possible role of a blocking high in impacting the heat wave, does provide support for the role of the tropics-wide IWVE in causing regular heat stress throughout the pre-monsoon summer season, every year, contiguously throughout the tropics. This is observed in the Indian subcontinent, which regularly experiences pre-monsoon heat stress (Pai et al., 2013), which could have a dominant component from the continuously varying humidity resulting in extremes in temperature.

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

#### 487 4 Conclusion

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

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

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

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

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

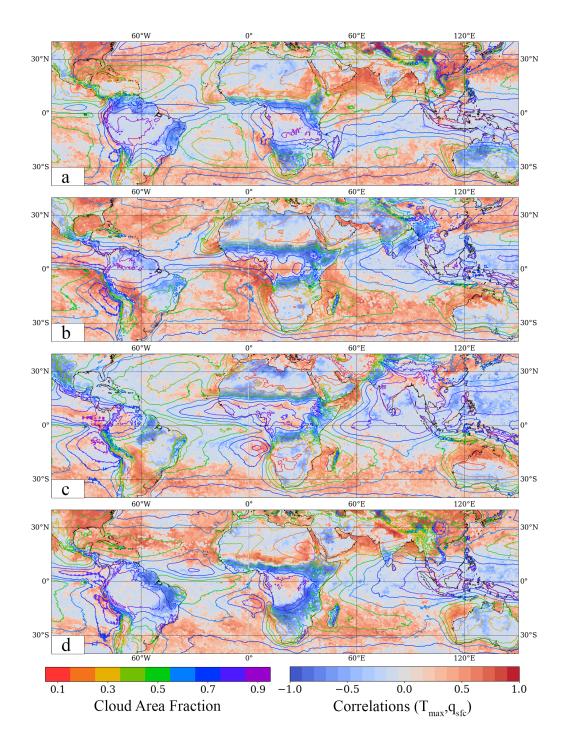


Figure 1. The tropical indirect water vapor negative radiative effect and its cooccurrence with transition regions in cloud fraction. Colored Shading - Mode of monthly daily scale correlations between daily  $T_{max}$  and  $q_{sfc}$  in (a) February, (b) May, (c) August, and (d) November. The mode was 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.

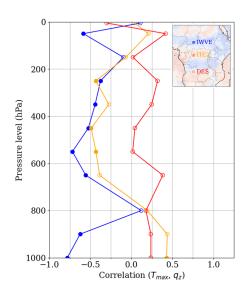


Figure 2. IWVE regions show negative correlations between  $T_{max}$  and upper tropospheric specific humidity. Correlation coefficients between daily  $T_{max}$  and daily average specific humidity at various altitudes,  $q_z$ , in May 2016 at selected IWVE, ITCZ and DES regions in Africa (inset). Data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

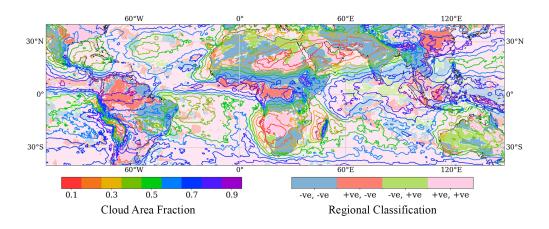


Figure 3. Dominant IWVE regions in the tropics. Classification of various regions into four types based on whether the sign of the grid scale correlations between daily  $T_{max}$  and  $q_{sfc}$  is the same as that between  $T_{max}$  and CWV. Area 1 - both negative, Area 2 - former positive, latter 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.

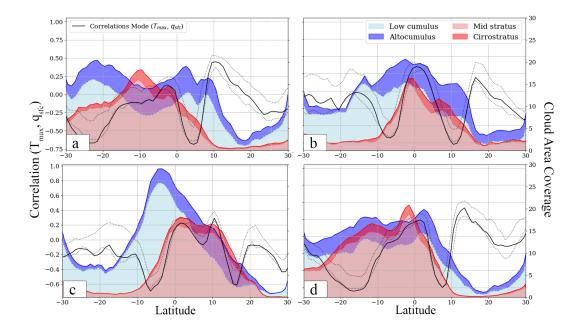


Figure 4. Dominant presence of mid level clouds in IWVE regions. Daytime cloud fraction over Africa, zonally averaged between  $30^{\circ}$ N,  $17.5^{\circ}$ E and  $30^{\circ}$ S,  $30^{\circ}$ E for (a) February, (b) May, (c) August, and (d) November, 2001 to 2020. Could types shown - Low cumulus including shallow cumulus (low, thin) and stratocumuls (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  $1^{\circ} \times 1^{\circ}$  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.

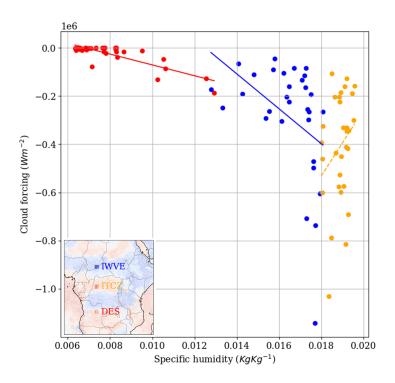


Figure 5. Surface cloud radiative forcing in IWVE regions is sensitive to near surface atmospheric humidity. Daytime averaged CRF versus  $q_{sfc}$  for a 1°×1° IWVE, ITCZ and a DES region in Africa (inset), for all days in May 2016. All data from ERA5.

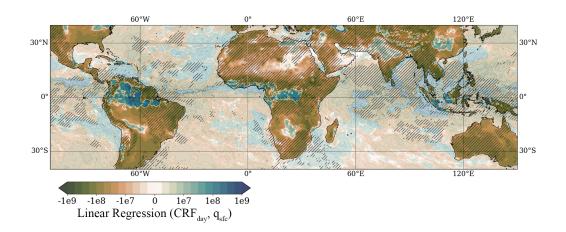


Figure 6. IWVE across tropics show a negative relation of surface humidity with cloud radiative forcing. Coefficients of linear regression of daytime CRF on daily averaged  $q_{sfc}$ . Regression performed with ERA5 grid scale, daily data for May 2016. Most values above -7e7 are significant at the 95% 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.

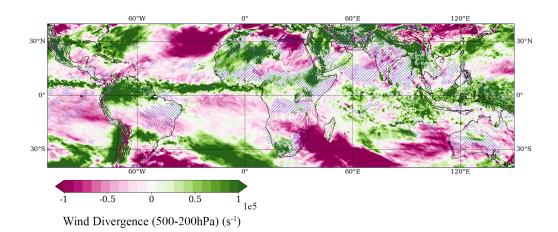


Figure 7. IWVE associated with the horizontal branch of continental scale overturning circulations. Shading shows the divergence of horizontal winds, column integrated between 500 hPa and 200 hPa averaged over May 2016. Hatching shows IWVE regions with  $T_{max}$ ,  $q_{sfc}$  correlation coefficients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

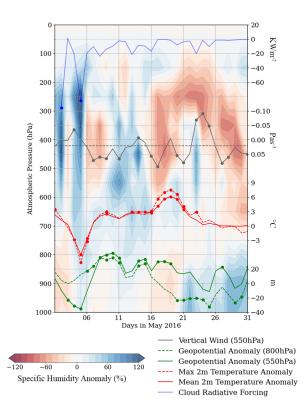


Figure 8. Severe tropical heat stress could be related to IWVE. Daily scale time series of  $T_{max}$ ,  $T_{avg}$  (daily mean),  $GH_{800}$  and  $GH_{550}$  anomalies from the climatological mean corresponding to May 1979 to 2020. Figure also shows the percentage temporal anomalies in  $q_z$  from the monthly mean of May 2016, the daytime CRF and daily 550 hPa vertical wind  $\omega_{550}$ . The dashed grey line represents the climatological average vertical wind in May. Significant anomalies are marked. Significant differences of CRF and  $\omega_{550}$  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,  $\Delta GH_{550}$ . 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  $\Delta T_{max}$  and  $\Delta T_{avg}$ , daily scale IWVE correlations corr $(T_{max},q_{sfc})$  and corr $(T_{max},CWV)$ , daily scale area averaged vertical wind at 550 hPa  $\omega_{550}$ , daytime averaged surface CRF, anomaly in the geopotential heights at the 800 hPa and 550 hPa levels during the heat wave period ( $\Delta GH_{800}$ ,  $\Delta GH_{550}$ ), and correlations corr $(T_{max},GH_{50})$ . Some of these values are also reported for the day of maximum  $\Delta GH_{550}$ . 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 (°N)	27.13	25.67	26.87	16.5	23.75	18.75
longitude (°E)	72.36	78.45	80.94	79.75	71.75	75.5
$\operatorname{avg}(T_{avg,May})$	34.96	34.05	33.78	33.2	33.7	32.38
Day of $\max(T_{max})$	19	18	17	24	18	17
$\Delta T_{max}$ (°C)	7.51	3.97	4.26	4.16	6.14	3.79
$\Delta T_{avg}$ (°C)	6.1	3.83	3.54	4.05	3.99	3.8
$\operatorname{corr}(T_{max}, q_{sfc})$	-0.4	-0.37	-0.47	-0.81	-0.37	-0.45
$\operatorname{corr}(T_{max}, CWV)$	-0.62	-0.41	-0.57	-0.67	-0.69	-0.24
$\omega_{550} (\operatorname{Pa} \operatorname{sec}^{-1})$	0.0	0.0	0.01	0.03	0.03	0.03
$CRF~(\mathrm{kWm^{-2}})$	-1.51	0	-19.1	-33.39	0.5	-18.97
$\Delta GH_{800}$ (m)	-13.16	-14.83	-11.04	3.86	-10.91	-20.2
$\Delta GH_{550}$ (m)	27.6	14.37	6.27	7.62	19.74	-2.96
$\operatorname{corr}(T_{max}, GH_{800})$	-0.2	-0.39	-0.17	0.33	0	-0.13
$\operatorname{corr}(T_{max}, GH_{550})$	0.38	0.17	0.32	0.39	0.4	0.31
Day of	10	14	14	14	14	14
$\max(\Delta GH_{550})$						
$\Delta T_{max}$ (°C)	2.27	2.73	3.11	1.01	2.43	2.38
$\Delta GH_{550}$ (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

### 526 5 Open Research

[H] The meteorological data used for the heat stress analysis in the study is avail-527 able freely on the Copernicus Climate Data Store via 10.24381/cds.143582cf, https:// 528 cds.climate.copernicus.eu/#!/search?text=ERA5&type=dataset (Hersbach et al., 529 2020, 2023). Supporting meteorological data from the AIRS L3 standard data products 530 V7 is obtained from the data collections at NASA GES DISC via 10.5067/UO3Q64CTTS1U, 531 https://disc.gsfc.nasa.gov/datasets?page=1&source=AQUA%20AIRS&keywords=airs% 532 20version%207 (AIRS, 2019). CERES cloud type and surface radiation flux data are 533 obtained from the SYN1deg level 3 products from the NASA CERES data repository via (doi - 10.5067/TERRA+AQUA/CERES/CLDTYPHIS\_L3.004, 535 doi - 10.5067/Terra+Aqua/CERES/SYN1degDay\_L3.004A respectively) https://ceres 536 .larc.nasa.gov/data/ (Doelling et al., 2013, 2016). All datasets used are available freely 537 for public use. 538

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.

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

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Key Points:

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9	An indirect, water vapor related surface cooling is observed at the daily scal	le, 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 sea	sonal

cycle, possibly consequential for pre-monsoon heat stress

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#### 15 Abstract

We report an indirect, negative, cloud-mediated, surface radiative effect (RE) of water 16 vapor (IWVE) in certain regions in the tropics, which may be consequential for day-to-17 day regional heat stress. Using reanalysis and satellite data we show that this effect is 18 marked by a surprisingly dominant positive relationship of cloud RE with near surface 19 and column humidity. These clouds are predominantly low level and altocumuli, previ-20 ously reported to have a negative surface RE, possibly lending the net negative RE to 21 water vapor. Also reported earlier, these clouds form in the mid-troposphere, as detrain-22 ment offshoots of deep convective towers and can be advected away to large distances, 23 hence requiring no local convective triggering in the IWVE regions. Evidently, the IWVE 24 are co-located with the horizontal branch of the Hadley cell, with the lowest vertical forc-25 ing in the tropics. Moreover, these are also the transition regions between the highly cloudy 26 and the driest parts of the tropics, with a waning down occurrence of cirrus, deep con-27 vective and altostratus clouds, linked with positive RE, corroborating the hypothesis. 28 IWVE regions also show a large temporal variability in humidity possibly providing op-29 portunity for a large variability in cloud fractional coverage, however the mechanism con-30 trolling this covariability is not understood. The IWVE is tightly tied with the seasonal 31 cycle of the ITCZ and hence is likely a dominant source of pre-monsoon surface temper-32 ature variability and heat stress in the current climate. The evolution of the IWVE un-33 der future climate warming needs further investigation. 34

#### <sup>35</sup> Plain Language Summary

The greenhouse radiative effect of water vapor on surface temperatures is widely 36 known. However, can water vapor cause an 'apparent' radiative cooling effect under cer-37 tain conditions? Here, we show that in certain extensive tropical regions, bordering the 38 ITCZ, the lower tropospheric water vapor has a substantial relation with cloud occur-39 rence, which subsequently renders an indirect surface cooling effect in high humidity con-40 ditions. These clouds are majorly shallow and mid level cumuli, which are known to have 41 a negative surface radiative forcing. What causes the positive relation between lower tro-42 pospheric humidity and these clouds, is not understood. However, that these regions bor-43 der the ITCZ and hence are co-located with the horizontal branch of the Hadley cell, sug-44 gests low amounts of convective triggering and hence high level clouds. The potentially 45 reduced warming from the high level clouds may give the lower level clouds an upper hand 46 in the net radiative impact. Lastly, these regions precede the ITCZ in time as they move 47 north-south with it. This may be consequential for pre-monsoonal heat stress in trop-48 ical regions. 49

#### 50 1 Introduction

Water vapor is frequently dubbed as the most important greenhouse gas which plays 51 a dominant role in climate change through the water vapor feedback (Soden et al., 2002, 52 2005; Forster et al., 2021; Dessler et al., 2013). This control is delivered through the dom-53 inant impact of the increase in the upper tropospheric humidity on trapping the outgo-54 ing longwave radiation (OLR) and recycling it within the climate system (Soden et al., 55 2005; Allan et al., 1999; Dessler et al., 2013). This trapping of energy warms up all com-56 ponents of the climate system - the atmosphere, land and oceans. However, on hourly 57 to daily time scales, it is the near surface humidity (or column integrated water vapor, 58 as they are strongly correlated) which dominates the greenhouse effect of water vapor 59 as manifested at the land surface (Shakespeare & Roderick, 2021; Flerchinger et al., 2009; 60 M. Li et al., 2018; Brutsaert, 1975; Carmona et al., 2014). Idealized modeling studies 61 of clear sky surface downwelling longwave radiation (DLR) are able to explain about 97%62 of the temporal variability in DLR with near surface humidity  $(q_{sfc})$  and surface tem-63 peratures  $(T_{sfc})$  (Shakespeare & Roderick, 2021). High positive correlations have been 64

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

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

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

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

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 124 with local cloud cover, on daily scale surface heat stress in tropical regions during the 125 peak, pre-monsoon, summer season. The general correlations between the humidity and 126 surface temperature variables are presented first, subsequently relating the observation 127 to the local cloud cover and variability. Next, the relationship of these correlations with 128 the larger scale atmospheric dynamics is presented. Lastly, a short analysis is presented 129 to propose a possible role of the near surface humidity in causing observed heat stress 130 through the processes identified. 131

#### 132 2 Methods

This study focuses on processes that impact the daily scale variability of regional 133 surface temperatures  $(T_{sfc})$ . As such, most analysis uses daily scale data for a selected 134 month. This choice of time scales reflects our interest in the 'instantaneous' response of 135  $T_{sfc}$  to the radiative forcing by the variability in water vapor and clouds. Since the lat-136 ter process takes of the order of a few hours to manifest, the day-to-day co-variability 137 in the variables is analyzed, if not the co-variability at sub-daily time scales. The atmo-138 spheric water vapor variability at longer time scales might itself be caused by inter-seasonal 139 and inter-annual climate variability and climate change, but that is not the focus of this 140 study. However, our analysis with inter-annual, monthly averaged variables results in sim-141 ilar inferences as the analysis with daily scale variables, providing support for the gen-142 eralizability of the processes identified. Results are majorly reported from the pre-monsoon 143 summer season of the northern hemisphere, but the inferences remain unchanged for other 144 seasons. The month of May 2016 is chosen for some specific analysis that is limited by 145 the size of the data set involved. 2016 was a post El Niño year transitioning to a mild 146 La Niña, infested by high heat stress in the Indian subcontinent (Oldenborgh et al., 2018). 147

Surface heat stress is quantified with daily scale maximum surface temperatures 148  $(T_{max})$  from reanalysis and satellite datasets.  $T_{max}$  appears frequently and takes an im-149 portant role in the formal estimation of heat stress (IMD, 2002; Perkins, 2015). How-150 ever, in this study, heat stress is used to refer to periods of statistically significant de-151 partures of  $T_{max}$  from the climatological or monthly mean and should not be confused 152 with the various heat wave indices used by meteorological departments to declare pe-153 riods of heat stress. Most inferences reported here with  $T_{max}$  remain unchanged for an 154 analysis with daily averaged surface temperatures  $(T_{avg})$  and daily maximum 2 m air tem-155 perature  $(T_{2m,max})$ . 156

The daily averaged near surface specific humidity  $(q_{sfc})$ , specific humidity at var-157 ious altitudinal levels  $(q_z)$  and total column water vapor (CWV) are used to capture the 158 radiative effect of atmospheric water vapor on surface temperatures. Previous studies 159 have utilized these variables to show that at daily to monthly timescales, in most global 160 land areas these variables are non-linearly but positively correlated with the surface down-161 welling longwave radiation (DLR) or surface temperatures (Gaffen et al., 1992; Ross et 162 al., 2002). Although the clear sky regression coefficients between  $q_{sfc}$  and DLR are space 163 dependent (Allan et al., 1999), there is no previous evidence of them being negative. Hence, 164  $q_{sfc}$  is used to represent the greenhouse impact of atmospheric water vapor on surface 165 temperatures (Shakespeare & Roderick, 2021). 166

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

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

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

Most analysis in this work is performed using statistics like linear correlation coefficients, linear regressions coefficients and anomalies from spatio-temporal averages for daily scale data analyzed over a month. Statistical significance is obtained with a twotailed t-test for correlation coefficients, linear regressions and differences at the 95% confidence level in all cases.

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

Figures 1 and S1 show the multiyear mode of monthly correlations of daily scale  $T_{max}$  (or  $T_{2m,max}$ ,  $T_{avg}$ ) and  $q_{sfc}$  (or CWV). Multiyear mode is calculated using gridscale monthly correlations for 20 years, from four neighboring grid points, regardless of the statistical significance of the correlations in individual years. However, the correlations of interest (IWVE correlations defined in 'Results and Discussion') are robustly significant across all years. Fig. S1b is shown as an example and will be discussed in detail in the next section.

#### <sup>209</sup> **3** Results and Discussion

The objective of this study is to identify regions where water vapor impacts surface heat stress through processes over and above its greenhouse warming impact. In other words, we endeavor to investigate regions where the direct greenhouse effect of water vapor may be masked by the radiative impacts of the other climatically important processes mediated by water vapor, like cloud formation. To identify such regions, a basic correlation analysis between daily scale surface temperature and indicators of atmospheric humidity is performed. 217 218

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

At the daily scale,  $T_{max}$  of a majority of regions around the globe are strongly pos-219 itively correlated with  $q_{sfc}$  (Fig. 1), and CWV (Fig. S1 a) in ERA5 and also in AIRS 220 (Fig. S2a). The values of these correlation coefficients for May 2016, are displayed in 221 Fig. S1 b to show that the spatial patterns of the multiyear mode of these correlations 222 (Figs. 1, S1) compare well with the values for individual years. These widespread pos-223 itive correlations could be attributed to the greenhouse effect of water vapor. The max-224 225 imum positive correlations larger than  $\approx 0.7$  are obtained at high latitudes and moderate values over the tropical oceans, a feature of the surface temperature-humidity rela-226 tionship reported before (Ross et al., 2002). The inhibition of these positive correlations 227 along the inter-tropical convergence zone (ITCZ) over land could be attributable to the 228 dominant presence of high level clouds in these regions, which may have a complex net 229 *CRF*. The analysis of these regions is out of the scope of this study. 230

The same figures, however, also show several tropical and sub-tropical land regions 231 with high, statistically significant, negative correlations between daily  $T_{max}$  and  $q_{sfc}$  which 232 occur across seasons and the two data sets used. Such signal has previously been observed 233 over western USA using radiosonde data (Ross et al., 2002). In our study the negative 234 correlation coefficients range between -0.36 to -0.97 for data averaged for May between 235 2001 and 2020 (Fig. 1). The largest negative correlations are found predominantly over 236 tropical Africa and tropical South America throughout the year and over the Indian sub-237 continent, southeast Asia and northern Australia in some seasons. Similar patterns ap-238 pear also with  $T_{avg}$  and  $T_{2m,max}$  (Figs. S1 c and S1 d) and with inter-annual correlations 239 using monthly averaged data (Fig. S2b), indicating the generalizability of this relation-240 ship. This apparent water vapor related cooling is observed over land areas on the flank 241 regions of the ITCZ and follows its seasonal cycle (Fig. 1). Hence, for example over the 242 Indian subcontinent, it comes into prominence in the dry, pre-monsoon summer season 243 (Fig. S3). It should be noted however that these negative correlations might maximize 244 in a region's peak monsoon period. This is probably due to an atmospheric moistening 245 and a sharp evaporative surface cooling due to heavy precipitation - for example in June 246 over India when the monsoons arrive over land (Fig. S3). However, the cause of the wa-247 ter vapor related cooling in the pre-monsoon dry period is not obvious. 248

This negative correlation of near surface water vapor with surface temperatures, 249 suggesting an apparent negative radiative forcing from water vapor is termed here as the 250 'Indirect Water Vapor Effect' or IWVE. This apparent indirect forcing of water vapor 251 is investigated next. In the following, IWVE correlations will mean a correlation between 252 daily  $q_{sfc}$  and  $T_{max}$ , unless otherwise specified. Furthermore, the positive IWVE cor-253 relation regions over the deep convective regions near the equator are termed as the ITCZ 254 and the positive IWVE correlation regions polewards of the IWVE regions are termed 255 as the desert or DES regions. 256

Figure 2 shows the correlations between specific humidity at an altitude  $(q_z)$  and 257 daily  $T_{max}$  over three regions of the African subcontinent - an IWVE, an ITCZ and a 258 DES region in May 2016. These regions are carefully chosen to lie at the same longitude 259 (19°E) and away from topography. Over the IWVE region, the correlation between  $T_{max}$ 260 and  $q_z$  is always significantly negative throughout the troposphere. Similarly, over the 261 DES region the above correlation is significantly positive throughout. However, over the 262 ITCZ region the correlation is positive close to the surface but changes to negative in 263 the upper half of the troposphere. The altitudinal dependence of the correlations over 264 265 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. 266 Only the IWVE regions, where  $q_{sfc}$  dominates the overall correlation between CWV and 267  $T_{max}$ , are studied here. This is a reasonable choice because most IWVE regions defined 268

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

271 272

# 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 273 indicated time periods from ERA5. The IWVE co-occurs with regions of transition cloud 274 cover, tropics-wide, i.e. between regions of maximum and minimum cloud cover where 275 its gradient is substantial. For example, during May over central Africa (Fig. 1b), the 276 average cloud cover over the IWVE regions is around 0.51 (standard deviation 0.17), as 277 compared to around 0.71 (0.18) over the ITCZ and 0.24 (0.18) over the DES regions. Hence, 278 the total cloud cover in the IWVE regions is intermediate between regions of predom-279 inant convection (ITCZ) and regions of predominant downwelling (over subtropical deserts 280 like the Sahara, Kalahari, the Arabian desert etc). Additionally, the fact that the IWVE 281 regions tightly border the equatorial regions of deep convection everywhere, and follow 282 the north-south march of the ITCZ, suggests that the IWVE correlations could be me-283 diated by the radiative effects of some specific clouds that are dominant at the borders 284 of the ITCZ. 285

Next, a possible pathway, mediated by clouds, is analyzed, which may lead to the 286 apparent negative radiative forcing from water vapor in the IWVE regions. These re-287 gions receive much lower precipitation than the ITCZ (Fig. S4) which suggests that the 288 IWVE regions may harbor clouds, or a cloud distribution, that is distinctly different from 289 those in the ITCZ, resulting in a distinctly different CRF. Clouds show a spectrum of 290 surface and top of atmosphere CRF depending upon their cloud top height and opti-291 cal depth (Chen et al., 2000; Hartmann et al., 1992). A radiative transfer calculation is 292 often necessary to obtain the net radiative effect of a certain vertical distribution of clouds. 293 However, in this study we have taken evidential support from previous research to tie 294 the observed cloud distribution over the IWVE regions with the negative, apparently cloud-295 mediated radiative impact of  $q_{sfc}$ , both of which are discussed below. A comprehensive 296 numerical analysis of these relationships is left for a follow up study. 297

298

#### 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 zon-299 ally averaged over a region in Africa  $(30^{\circ}N, 17.5^{\circ}E \text{ and } 30^{\circ}S, 30^{\circ}E)$ , obtained from the 300 CERES CldTypHist product for May 2001 to 2020. The corresponding May average cloud 301 maps for Africa are provided in Figure S5. There is a clear distinction in the distribu-302 tion 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 mod-304 erately thick clouds, like shallow cumuli, altocumuli and stratocumuli. While the ITCZ, 305 apart from the low and mid level clouds, also have a higher percentage of optically thin 306 to thick, mid and high clouds, relative to the IWVE regions. This feature is present in 307 all seasons except when the IWVE correlations are large during the monsoon, apparently 308 due to the presence of strong evaporative surface cooling (August in Northern Hemisphere 309 in Figure 4c), in which case the low and altocumulus clouds become infrequent over the 310 IWVE regions. Specifically, the non-monsoonal, IWVE regions show a dominant pres-311 ence of mid level, low thickness altocumulus clouds, which follow the poleward bound-312 aries of the IWVE regions and are present abundantly and uniformly between these lat-313 itudes (Figure S5d). 314

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 (opti-319 cal depth around 1.03) mid level clouds (3 to 8 km) with a possibly negative radiative 320 forcing over land has been previously reported in the pre-monsoon months over west-321 ern Africa in the Nigerian region (Bouniol et al., 2012; Bourgeois et al., 2016; Stein et 322 al., 2011), in Darwin (Riihimaki et al., 2012) and in the pre-monsoon Indian subconti-323 nent (Parker et al., 2016). These mid level, thin, altocumulus clouds with a negative ra-324 diative forcing in the tropical regions (Bourgeois et al., 2016), are associated with mid 325 level detrainment from deep convective clouds near the zero degree isotherm (Johnson 326 et al., 1996, 1999; Iwasa et al., 2012). It is suggested that the latent heat of condensa-327 tion to ice causes the deep convective tower to lose buoyancy and form these clouds, which 328 can get horizontally advected to far away regions. As such, the occurrence and radia-329 tive impact of such mid level clouds is likely to be important in the flank regions of the 330 equatorial deep convective belt, which is where the IWVE regions occur. 331

Hence, the dominant shallow- and alto- cumulus clouds in the IWVE regions could 332 be responsible for the observed surface cooling. However, the question - how is  $q_{sfc}$  in 333 the IWVE regions linked to surface cooling caused through these clouds? - is still open. 334 This is an important question, because the expected positive, greenhouse radiative forc-335 ing of water vapor seems to be negated by a mechanism that supports clouds with a neg-336 ative radiative impact. This mechanism should cause the aforementioned cooling clouds 337 to occur along with high  $q_{sfc}$  in the IWVE regions. Given the strength and robustness 338 of the IWVE correlations over several time and spatial scales, this mechanism could have 339 great significance for the heat stress dynamics in the IWVE regions. Although the ex-340 act mechanism is not investigated here, we show evidence of such a negative relation-341 ship between CRF and  $q_{sfc}$  that exists dominantly over the IWVE regions, rendering 342 the 'effective' negative radiative forcing from water vapor. 343

Over the IWVE regions, we find a robust, significant and large negative (positive) 344 regression between  $q_{sfc}$  and daytime CRF (cloud fractional coverage, figure not shown), 345 but insignificant regression between these quantities in the DES regions and positive, in-346 significant values in some areas of the ITCZ (Fig. 5). This feature of the IWVE, ITCZ 347 and DES regions is observed pan-tropics (Figures 6 and S6) throughout the year (fig-348 ures not shown). However, we do find the ITCZ regions to have a significant negative 349 correlation between upper tropospheric water vapor and CRF (figure not shown) indi-350 cating a dependence of cloud formation on local environmental humidity, as is presum-351 ably also the case over the IWVE regions. In the IWVE regions, we however find this 352 strong dependence of CRF on the water vapor present throughout the column (figure 353 not shown). In other words, in the IWVE regions, water vapor in the lower and higher 354 tropospheric levels seems to vary and impact cloud cover in tandem. The mechanism be-355 hind this apparent correlation is not understood, however, the correlation enables lower 356 tropospheric water content to be a good indicator of CRF in these regions. 357

Lastly, the IWVE regions also show a large day-to-day variability in both cloud cover 358 and CRF, which is not present in the DES and ITCZ regions (Fig. 5). Over DES,  $q_{sfc}$ 359 is too low to be associated with substantial cloud occurrence, whereas, over the ITCZ 360 it is too high to limit or control the same. While we have not established any cause-and-361 effect relationships between  $q_{sfc}$  and cloud cover in the three regions, the general atmo-362 spheric conditions do seem to be less conducive to or supportive of clouds over the DES 363 regions and vice versa over the ITCZ. Some of this possibly could be attributed to the 364 stable conditions that form over the DES regions due to Hadley circulation and vice versa 365 over the ITCZ. The IWVE regions however show a large variability in both  $q_{sfc}$  and CRF. 366 suggesting atmospheric conditions variable enough to support different types and amounts 367 of clouds. This feature is observed in most IWVE regions throughout the tropics (fig-368 ure not shown). The effect of tropical latitudinal overturning circulation on the IWVE 369 regions is analyzed next. 370

371 372

## 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 380 the year due to maximum solar insolation, which results in the high cloud cover and pre-381 cipitation associated with these regions discussed earlier. The divergence of the horizon-382 tal wind, column integrated between 500 hPa and 200 hPa in the tropics, shows positive 383 values (divergence) over the ITCZ regions (Fig. 7). Hence, the regions where  $q_{sfc}$  is pos-384 itively (not) correlated with  $T_{max}$  (CRF) are also the regions of deep convective forc-385 ing. The high cloud cover of all types in these regions supports this inference (Fig. 4). 386 The opposite is true for DES regions which show predominant horizontal convergence, 387 downwelling and clear skies (Fig. 7). 388

The IWVE regions, on the other hand, are marked by the smallest horizontal di-389 vergence and vertical motion as depicted in Figure 7. This is seen by the overlap between 390 the IWVE regions and the contours of near zero divergence of the horizontal winds in-391 tegrated over the upper troposphere. This feature is present over all IWVE regions in 392 the tropics. This absence of a strong vertical forcing could be related with the dominance 393 of shallow cumulus and altocumulus clouds and the lower areal coverage of deep convec-394 tive and cirrus clouds in the IWVE regions (Figure 4). It can only be speculated now 395 and verified in a numerical study that vertical forcing is conducive to a certain type of 396 clouds, the absence of which may be related with the IWVE. 397

398

#### 3.5 Why could the water vapor indirect effect be important?

The co-location of the IWVE regions with transition cloud cover and negligible ver-399 tical motion (or convective forcing) indicate that these regions occur in the horizontal 400 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 402 find a strict latitudinal dependence of the IWVE regions over various continents in a given 403 season. In fact, continental scale overturning circulations, like the one found over the north-404 ern Sahara in May 2016 (Figure 7), can also engender IWVE regions co-located with their 405 horizontal branches. However, Figure 1 demonstrates clearly that the IWVE is not present 406 in regions between Libya and Mali over a climatological average and may not be impor-407 tant for the regional heat stress climatology. However, in regions affected by a climato-408 logical latitudinal overturning circulation, like the Hadley cell, the IWVE could be de-409 cisive in determining pre-monsoonal surface heat stress. 410

To this effect, we provide an example of the surface heat stress during the severe, 411 pan-India heat wave of May 2016. An exhaustive analysis of the general heat stress at 412 these latitudes, their relation with the IWVE and any deviations will be presented in a 413 follow up study. India lies in the Southeast Asian belt where both high temperature and 414 high humidity contribute significantly to heat stress (Im et al., 2017; Hoegh-Guldberg 415 et al., 2018; Wehner et al., 2016) particularly in April, May and June. While IWVE shows 416 417 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 Hyder-418 abad, India. Hence, this region has been chosen for a case study. Record high near sur-419 face temperatures of 51 degrees were observed in Phalodi, Rajasthan (27.13°N, 72.36°E) 420 around  $19^{th}$  May 2016 with the heat wave impacting regions pan-India between the months 421

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

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

Figure 8 shows the May 2016 time series of several meteorological variables over 443 Phalodi, Rajasthan, India. The variables  $q_z$ , pressure velocity at 550 hPa ( $\omega_{550}$ ) and CRF444 derived from ERA5, averaged over a  $2^{\circ} \times 2^{\circ}$  region centered around Phalodi are used to 445 understand the IWVE. Following studies of extra-tropical heat waves, the geopotential 446 height anomalies from a climatological mean (1979 to 2020) at 800 hPa and 550 hPa ( $\Delta GH_{800}$ ) 447  $\Delta GH_{550}$ ) are used to represent the role of blocking highs in triggering heat stress. Fig-448 ure S7 shows the same variables over other regions in the subcontinent during the same 449 heat wave. These regions were chosen because they have either a high IWVE,  $\Delta T_{max}$ 450 or  $\Delta GH_{550}$  signal. These figures are summarised in Table 1. 451

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

The above preliminary analysis suggests that statistically significant geopotential anomalies are not co-eval with the period of highest heat stress at least over the Indian subcontinent during the severe heat wave of May 2016. In fact, these periods may be separated from each other by several days. At the same time, every station is found to be significantly impacted by the IWVE, especially during extremes in surface temperatures. This observation, although does not preclude the possible role of a blocking high in impacting the heat wave, does provide support for the role of the tropics-wide IWVE in causing regular heat stress throughout the pre-monsoon summer season, every year, contiguously throughout the tropics. This is observed in the Indian subcontinent, which regularly experiences pre-monsoon heat stress (Pai et al., 2013), which could have a dominant component from the continuously varying humidity resulting in extremes in temperature.

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

#### 487 4 Conclusion

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

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

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

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

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

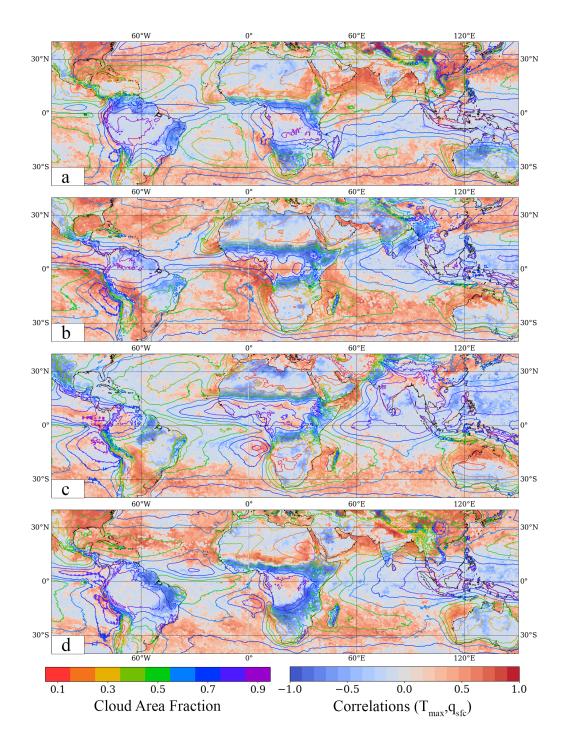


Figure 1. The tropical indirect water vapor negative radiative effect and its cooccurrence with transition regions in cloud fraction. Colored Shading - Mode of monthly daily scale correlations between daily  $T_{max}$  and  $q_{sfc}$  in (a) February, (b) May, (c) August, and (d) November. The mode was 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.

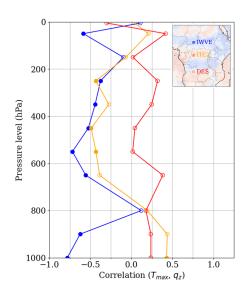


Figure 2. IWVE regions show negative correlations between  $T_{max}$  and upper tropospheric specific humidity. Correlation coefficients between daily  $T_{max}$  and daily average specific humidity at various altitudes,  $q_z$ , in May 2016 at selected IWVE, ITCZ and DES regions in Africa (inset). Data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

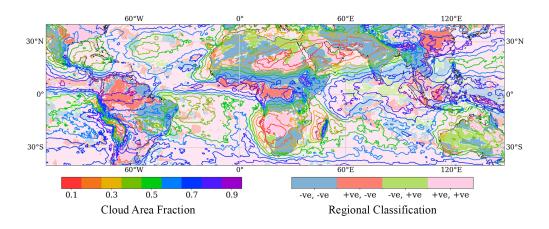


Figure 3. Dominant IWVE regions in the tropics. Classification of various regions into four types based on whether the sign of the grid scale correlations between daily  $T_{max}$  and  $q_{sfc}$  is the same as that between  $T_{max}$  and CWV. Area 1 - both negative, Area 2 - former positive, latter 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.

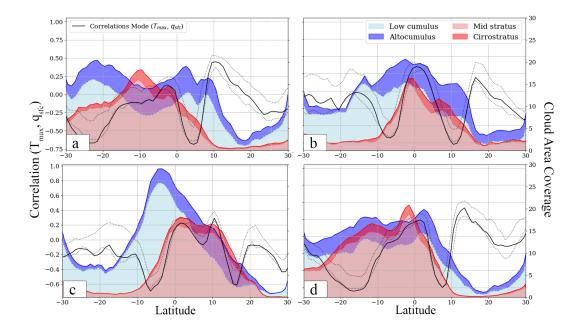


Figure 4. Dominant presence of mid level clouds in IWVE regions. Daytime cloud fraction over Africa, zonally averaged between  $30^{\circ}$ N,  $17.5^{\circ}$ E and  $30^{\circ}$ S,  $30^{\circ}$ E for (a) February, (b) May, (c) August, and (d) November, 2001 to 2020. Could types shown - Low cumulus including shallow cumulus (low, thin) and stratocumuls (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  $1^{\circ} \times 1^{\circ}$  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.

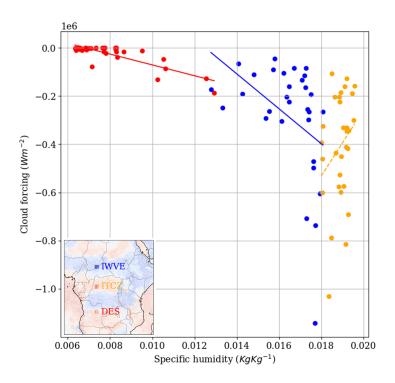


Figure 5. Surface cloud radiative forcing in IWVE regions is sensitive to near surface atmospheric humidity. Daytime averaged CRF versus  $q_{sfc}$  for a 1°×1° IWVE, ITCZ and a DES region in Africa (inset), for all days in May 2016. All data from ERA5.

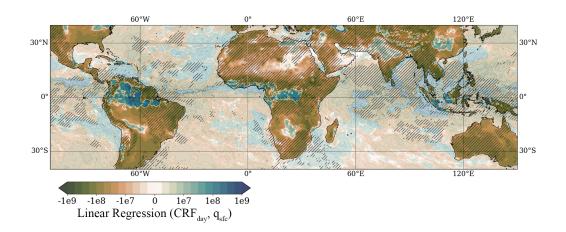


Figure 6. IWVE across tropics show a negative relation of surface humidity with cloud radiative forcing. Coefficients of linear regression of daytime CRF on daily averaged  $q_{sfc}$ . Regression performed with ERA5 grid scale, daily data for May 2016. Most values above -7e7 are significant at the 95% 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.

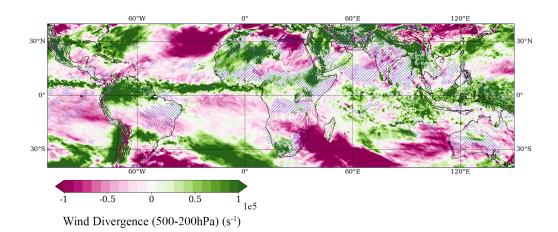


Figure 7. IWVE associated with the horizontal branch of continental scale overturning circulations. Shading shows the divergence of horizontal winds, column integrated between 500 hPa and 200 hPa averaged over May 2016. Hatching shows IWVE regions with  $T_{max}$ ,  $q_{sfc}$  correlation coefficients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

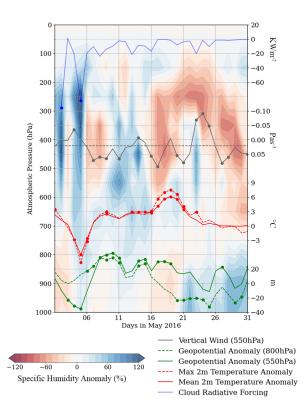


Figure 8. Severe tropical heat stress could be related to IWVE. Daily scale time series of  $T_{max}$ ,  $T_{avg}$  (daily mean),  $GH_{800}$  and  $GH_{550}$  anomalies from the climatological mean corresponding to May 1979 to 2020. Figure also shows the percentage temporal anomalies in  $q_z$  from the monthly mean of May 2016, the daytime CRF and daily 550 hPa vertical wind  $\omega_{550}$ . The dashed grey line represents the climatological average vertical wind in May. Significant anomalies are marked. Significant differences of CRF and  $\omega_{550}$  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,  $\Delta GH_{550}$ . 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  $\Delta T_{max}$  and  $\Delta T_{avg}$ , daily scale IWVE correlations corr $(T_{max},q_{sfc})$  and corr $(T_{max},CWV)$ , daily scale area averaged vertical wind at 550 hPa  $\omega_{550}$ , daytime averaged surface CRF, anomaly in the geopotential heights at the 800 hPa and 550 hPa levels during the heat wave period ( $\Delta GH_{800}$ ,  $\Delta GH_{550}$ ), and correlations corr $(T_{max},GH_{50})$ . Some of these values are also reported for the day of maximum  $\Delta GH_{550}$ . 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 (°N)	27.13	25.67	26.87	16.5	23.75	18.75
longitude (°E)	72.36	78.45	80.94	79.75	71.75	75.5
$\operatorname{avg}(T_{avg,May})$	34.96	34.05	33.78	33.2	33.7	32.38
Day of $\max(T_{max})$	19	18	17	24	18	17
$\Delta T_{max}$ (°C)	7.51	3.97	4.26	4.16	6.14	3.79
$\Delta T_{avg}$ (°C)	6.1	3.83	3.54	4.05	3.99	3.8
$\operatorname{corr}(T_{max}, q_{sfc})$	-0.4	-0.37	-0.47	-0.81	-0.37	-0.45
$\operatorname{corr}(T_{max}, CWV)$	-0.62	-0.41	-0.57	-0.67	-0.69	-0.24
$\omega_{550} (\operatorname{Pa} \operatorname{sec}^{-1})$	0.0	0.0	0.01	0.03	0.03	0.03
$CRF~(\mathrm{kWm^{-2}})$	-1.51	0	-19.1	-33.39	0.5	-18.97
$\Delta GH_{800}$ (m)	-13.16	-14.83	-11.04	3.86	-10.91	-20.2
$\Delta GH_{550}$ (m)	27.6	14.37	6.27	7.62	19.74	-2.96
$\operatorname{corr}(T_{max}, GH_{800})$	-0.2	-0.39	-0.17	0.33	0	-0.13
$\operatorname{corr}(T_{max}, GH_{550})$	0.38	0.17	0.32	0.39	0.4	0.31
Day of	10	14	14	14	14	14
$\max(\Delta GH_{550})$						
$\Delta T_{max}$ (°C)	2.27	2.73	3.11	1.01	2.43	2.38
$\Delta GH_{550}$ (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

### 526 5 Open Research

[H] The meteorological data used for the heat stress analysis in the study is avail-527 able freely on the Copernicus Climate Data Store via 10.24381/cds.143582cf, https:// 528 cds.climate.copernicus.eu/#!/search?text=ERA5&type=dataset (Hersbach et al., 529 2020, 2023). Supporting meteorological data from the AIRS L3 standard data products 530 V7 is obtained from the data collections at NASA GES DISC via 10.5067/UO3Q64CTTS1U, 531 https://disc.gsfc.nasa.gov/datasets?page=1&source=AQUA%20AIRS&keywords=airs% 532 20version%207 (AIRS, 2019). CERES cloud type and surface radiation flux data are 533 obtained from the SYN1deg level 3 products from the NASA CERES data repository via (doi - 10.5067/TERRA+AQUA/CERES/CLDTYPHIS\_L3.004, 535 doi - 10.5067/Terra+Aqua/CERES/SYN1degDay\_L3.004A respectively) https://ceres 536 .larc.nasa.gov/data/ (Doelling et al., 2013, 2016). All datasets used are available freely 537 for public use. 538

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.

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# Supporting Information for "An indirect, negative radiative effect of water vapor in the tropics and its implications for regional surface heat stress"

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1. Figures S1 to S7

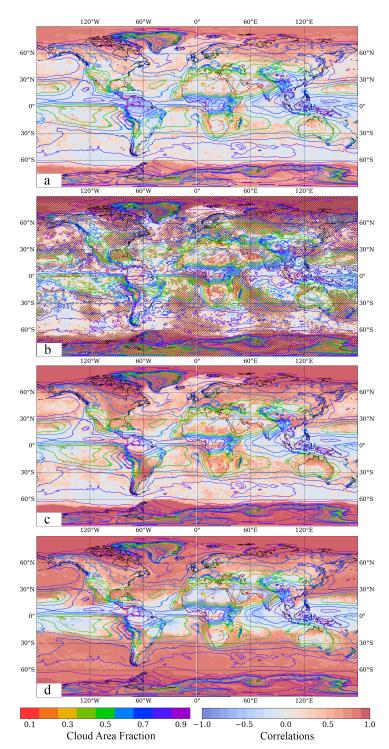


Figure S1. Correlations between – (a) daily  $T_{max}$  and CWV, (b) daily  $T_{max}$  and  $q_{sfc}$  for May 2016, (c) daily  $T_{avg}$  and  $q_{sfc}$ , and (d) daily  $T_{2m,max}$  and  $q_{sfc}$ . All data from ERA5. All coefficients are multi-year modes for May 2001 to 2020 (as in Fig. 1), except panel b which shows coefficients for May 2016. Contour lines represent cloud cover averaged over the corresponding period. Correlations significant at the 95% confidence level are hatched in panel b. September 24, 2023, 8:05am

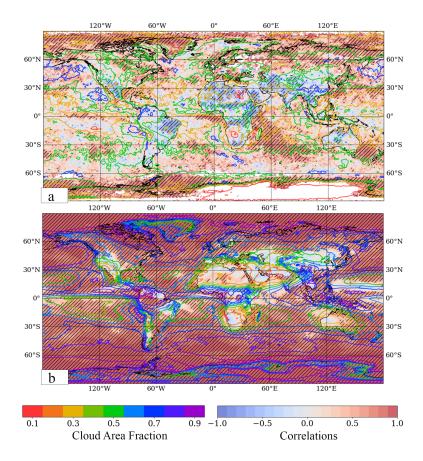


Figure S2. (a) Correlations between daily  $T_{max}$  and  $q_{sfc}$ , and daily averaged total cloud fraction from AIRS for May 2016, (b) Inter-annual correlations between monthly averaged  $T_{max}$  and  $q_{sfc}$ and corresponding daily averaged cloud cover for May 2001 to 2020 from ERA5. Correlations significant at the 95% confidence level are hatched.

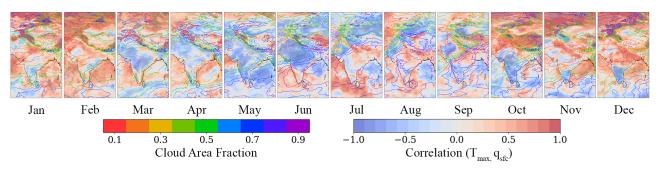


Figure S3. Grid scale correlations between daily  $T_{max}$  and  $q_{sfc}$  showing the seasonal pattern of IWVE over the Indian subcontinent for the year 2016.

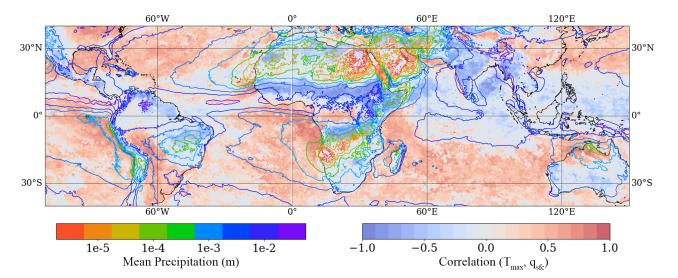
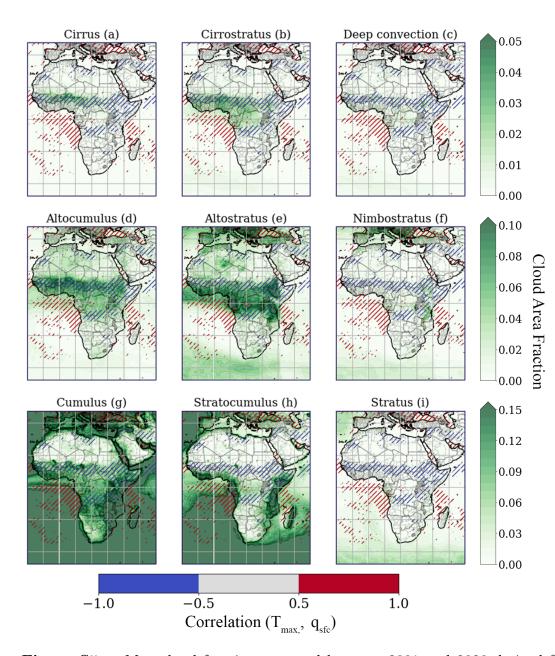
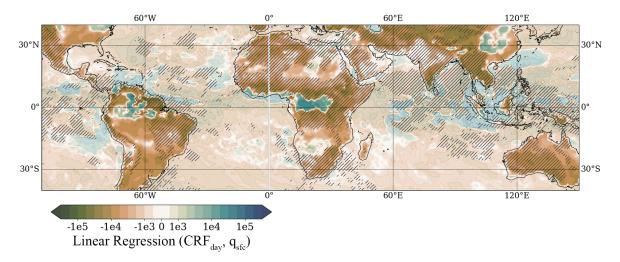


Figure S4. Same as Fig. 1 b but line contours show ERA5 precipitation averaged in May 2001 to 2020, instead of cloud fraction.



**Figure S5.** May cloud fraction averaged between 2001 and 2020 derived from CERES Cld-TypHist 1°x1° monthly data. Panels are arranged with highest clouds at the top and lowest clouds at the bottom. Optical depth increases from left to right. (a) Cirrus, (b) Cirrostratus, (c) Deep Convective clouds (d) Altocumulus, (e) Altostratus and (f) Nimbostratus, (g) Cumulus, (h) Stratocumulus and (i) Stratus. The blue and red hatching show regions with IWVE correlations between -1 to -0.5 and 0.5 to 1 respectively from Fig. 1b.



**Figure S6.** Same as Fig. 6 but with daily average *CRF* from CERES SYN1deg daily scale data for May 2016. Hatching shows all regions with negative IWVE correlations. Data obtained from ERA5.

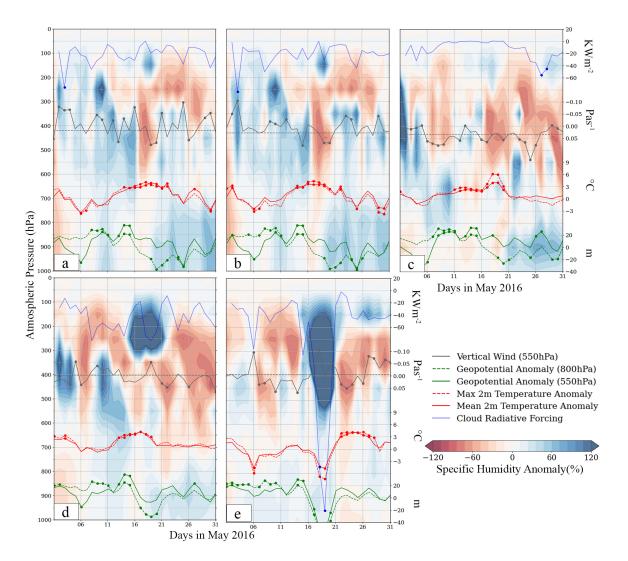


Figure S7. Same as Figure 8 but for (a) Datia, Madhya Pradesh (b) Lucknow, Uttar Pradesh,(c) Radhanpur, Gujarat, (d) Patoda, Maharashtra, and (e) Dachepalli, Andhra Pradesh.