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Process-Level Assessment of the Iris Effect Over Tropical Oceans

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Key Points:

- A-Train satellite observations are analyzed to investigate the physical processes considered to account for the iris effect
- The anvil cloud fraction reduces with increasing upper-tropospheric stability while unlikely related to precipitation efficiency
- The cloud radiative effects associated with the stability iris effect are nearly neutral when integrated over a diurnal cycle

Supporting Information:

Supporting Information may be found in the online version of this article.

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Abstract The iris hypothesis suggests a cloud feedback mechanism that a reduction in the tropical anvil cloud fraction (CF) in a warmer climate may act to mitigate the warming by enhanced outgoing longwave radiation. Two different physical processes, one involving precipitation efficiency and the other focusing on upper-tropospheric stability, have been argued in the literature to be responsible for the iris effect. In this study, A-Train observations and reanalysis data are analyzed to assess these two processes. Major findings are as follows: (a) the anvil CF changes evidently with upper-tropospheric stability as expected from the stability iris theory, (b) precipitation efficiency is unlikely to have control on the anvil CF but is related to mid- and low-level CFs, and (c) the day and nighttime cloud radiative effects are expected to largely cancel out when integrated over a diurnal cycle, suggesting a neutral cloud feedback.

Plain Language Summary Tropical anvil clouds, or extensive high clouds produced by deep convection, are known to be a key player in modulating the Earth's radiation budget. The iris hypothesis claims that anvil clouds have a stabilizing effect on the climate as they shrink and allow more heat to radiate out to space as the climate warms. The iris hypothesis, however, remains controversial partly because the processes behind it have not been well validated against observations. We analyze satellite observations and reanalysis data to test two known theories on the processes explaining the iris effect. The results show that a theory focusing on the air temperature structure around anvil clouds is likely at work in the tropical atmosphere, although the anvil's warming and cooling effects would offset each other during the whole day and night.

1. Introduction

Cirrus clouds prevail over tropical oceans, modulating the Earth's energy budget through the reflection of solar radiation and the absorption and emission of longwave radiation. Of particular interest are anvil cirrus clouds produced by detrainment from deep convection, which are a key player in the potential interactions between tropical convective dynamics and cloud radiative effects (CREs). There is rich literature on the cloud feedback involving the interactions of convection, clouds, and radiation. Ramanathan and Collins (1991) argued that the shielding of solar radiation by anvil clouds overwhelm the enhanced greenhouse effect in a moist, convectively active atmosphere over warm ocean, acting as a natural thermostat regulating sea surface temperature (SST). Other hypotheses put focus on the longwave effect of anvil clouds. Lindzen et al. (2001) proposed an "iris" mechanism that anvil clouds shrink and allow more warm radiation to escape out to space as sea surface temperature (SST) increases, resulting in a negative feedback on sea surface temperature (SST). The thermostat and iris hypotheses, although very different in mechanism and each challenged by a series of criticisms (see Pierrehumbert (1995) for discussions on the thermostat hypothesis and Hartmann and Michelsen (2002) and Lindzen and Choi (2021) on the iris effect), share a common ground in that anvil clouds play a role in stabilizing the climate system.

Another line of research (Hartmann & Larson, 2002; Zelinka & Hartmann, 2010) explored the possibility that the longwave effect of anvil clouds could give rise to a positive feedback rather than a stabilizing one. In the hypotheses known as fixed anvil temperature (FAT) and proportionately higher anvil temperature (PHAT), anvil cloud temperature remains nearly constant regardless of a surface warming, so that outgoing longwave radiation cannot efficiently remove an extra heat associated with the warming. The fixed anvil temperature FAT/PHAT theory is built upon an upper-tropospheric thermodynamic and mass balance consideration, interrelating a chain of processes such as temperature-limited moisture, radiative cooling, static stability, and horizontal divergence/convergence. Elements of this idea were reorganized into a "stability iris" theory (Bony et al., 2016) in light of a growing interest in convective self-aggregation (see reviews by Holloway et al., 2017; Wing et al., 2017). The

stability iris hypothesis predicts that anvil cloud cover should be reduced as the result of an enhanced upper-tropospheric stability in a warmer climate. The stability iris resembles the iris hypothesis as originally devised by Lindzen et al. (2001), whereas the underlying mechanism is entirely renewed.

Mauritsen and Stevens (2015) shed new light on the iris effect in the context of climate and hydrological sensitivities, demonstrating that climate model simulations are improved if the conversion from cloud water to precipitation is tuned so as to accelerate with a surface warming. Such a temperature dependence of precipitation efficiency was speculated by Lindzen et al. (2001) as a possible driver of the iris effect, although not supported to date by firm evidence (Sui et al., 2020). Rapp et al. (2005), Lin et al. (2006), and Choi et al. (2017) each evaluated precipitation efficiency in different manners and reached mixed conclusions regarding the iris effect. Observational evidence supportive of the stability iris was presented by Saint-Lu et al. (2020), although the consequences on cloud fraction (CREs) have yet to be examined.

A primary motivation of the present paper is to investigate the physical processes underlying the iris effect, if the iris exists in the real atmosphere. We do not seek observational evidence for the feedback hypothesis itself, given the difficulty of assessing the long-term regulation of SST with short-term observations. Our focus is restricted to the process-level linkage of precipitation efficiency or stability with the anvil cloud cover and cloud fraction (CREs).

At the heart of the analysis lie the measurements from the CloudSat radar and the Cloud-Aerosol Lidar and Infrared Pathfinder Satellite Observation (CALIPSO) lidar to sample anvil clouds and their parent convective clouds at the same time. This approach enables to preclude in-situ cirrus without direct relevance to convective dynamics, which amount to more than a half of the whole tropical cirrus (Luo & Rossow, 2004). A-Train satellite measurements and the European Centre for Medium-Range Weather Forecasts Reanalysis version 5 (ERA5) data are combined to evaluate the ratio of cloud water to precipitation (R_{CP}), introduced as a proxy of precipitation efficiency, and upper-tropospheric stability (S_{UT}). A composite analysis is then carried out to explore the dependence of anvil cloud properties on R_{CP} and S_{UT}. The net radiative effect is analyzed as well as the longwave effect for examining to what extent the shortwave component counteracts the longwave iris effect.

2. Data and Method

The Aqua satellite carries six instruments including Advanced Microwave Scanning Radiometer for the Earth Observing System (AMSR-E), Clouds and the Earth's Radiant Energy System (CERES), and Moderate Resolution Imaging Spectroradiometer (MODIS). The Aqua, CloudSat, and Cloud-Aerosol Lidar and Infrared Pathfinder Satellite Observation (CALIPSO) satellites, as part of the A-Train constellation, make near-simultaneous observations. The Cloud-Aerosol Lidar and Infrared Pathfinder Satellite Observation (CALIPSO), CloudSat, Clouds and the Earth's Radiant Energy System (CERES), and MODIS (CCCM) Release D1 product integrates observations from these instruments to yield the vertical structure of clouds and the in-cloud and clear-sky radiative flux estimates (Kato et al., 2011). The CCCM algorithm defines cloud fraction (CF) as the fractional coverage of CloudSat- and CALIPSO-detected clouds within each nadir ∼20-km Clouds and the Earth's Radiant Energy System (CERES) footprint. The in-cloud radiative flux is computed with the cloud optical properties diagnosed from the CloudSat, CALIPSO, and MODIS products combined together. The Advanced Microwave Scanning Radiometer for the Earth Observing System (AMSR-E) daily (practically instantaneous) product by Remote Sensing Systems (RSS) provide precipitation and liquid water path (LWP) estimates (Hilburn & Wentz, 2008; Wentz & Meissner, 2000). Upper-tropospheric stability and horizontal divergence are quantified using the European Centre for Medium-Range Weather Forecasts Reanalysis version 5 (ERA5) data (Hersbach et al., 2020).

The CCCM cloud and radiative properties are accumulated on a quarter-degree grid to match the Advanced Microwave Scanning Radiometer for the Earth Observing System (AMSR-E) and European Centre for Medium-Range Weather Forecasts Reanalysis version 5 (ERA5) datasets. Only the portions of AMSR-E and ERA5 data intersected by CloudSat/CALIPSO ground tracks are sampled for the analysis. Deep convection is defined to occur when cloud fraction (CF) exceeds 0.5 at all levels between 1 and 16.6 km in altitude for a given CCCM column. The lower limit of 1 km is chosen to be above the lifting condensation level as a proxy of convective cloud base, which is typically about 500 m above surface over tropical oceans. The upper boundary is set to be 16.6 km because the cold point tropopause reaches as high as 16–17 km in the Tropics (Gettelman & de F. Forster, 2002) and tropical deep convection often reaches close to this level (Takahashi & Luo, 2012).

In case that a sequence of CCCM pixels in row are identified as deep convection, the consecutive pixels are combined into a single convective event with their midpoint defined to be the geographical center of deep convection. It has been confirmed with sensitivity studies (not shown) that lower or higher cloud fraction (CF) thresholds than 0.5 do not qualitatively alter main conclusions, except that the sample size increases as the CF threshold is lowered. The cloud properties and atmospheric profiles within ±5° along CloudSat/CALIPSO tracks around each convection center are recorded for the composite analysis described next.

CloudSat and CALIPSO see only a limited portion of extensive cirrus clouds on an instantaneous basis. Nevertheless, "undersampled" convective events eventually constitute a large number of unbiased samples when accumulated over a period of time, given that satellite tracks homogeneously cover the globe and randomly intersect with cloud systems. The composite cloud structure shown in this work is thus considered to be a robust statistical representation of cirrus clouds in their entirety.

Composite analysis is carried out to look into the statistical characteristics of deep convection and the associated cloud properties and atmospheric states. Satellite measurements sampled around the deep convection centers are averaged together into a two-dimensional composite space comprised of along-track distance from the convection center and altitude. The compositing method is similar to the technique devised by Igel et al. (2014) but is substantially simplified. The primary interest of Igel et al. (2014) lay in the anvil cloud structure and how it changes with SST beneath, while the present work is focused more on the resulting radiative effects in reference to R_{CP} and S_{UT} (precise definitions are given later). To this end, composite samples are broken down by quartiles of R_{CP} and of S_{UT}, with outliers (i.e., below the first quartile minus 1.5 × IQR or above the third quartile plus 1.5 × IQR, where IQR is an abbreviation of the interquartile range) removed from the samples.

Precipitation efficiency is conventionally defined by the ratio of precipitation to either the large-scale moisture influx or the condensation rate (Sui et al., 2007). The variables appearing in the denominator, however, are difficult to evaluate from satellite observations alone. As a workaround, precipitation efficiency can be substituted by a combination of cloud water mass (or cloud size) and precipitation, both of which are available from satellite measurements (e.g., Lau & Wu, 2003; Rapp et al., 2005). In the present work, the ratio of precipitation P to column-integrated cloud water,

$$R_{CP} \equiv \frac{P}{\mathrm{LWP} + \mathrm{IWP}}, \tag{1}$$

is employed as a proxy of precipitation efficiency in the current analysis. Here liquid water path (LWP) and precipitation are obtained from the AMSR-E product, while ice water path (IWP) is estimated as the vertical integral of CCCM cloud ice water content. This compromise is necessitated by the technical limitations in satellite measurement capabilities that (1) microwave radiometry is insensitive to ice clouds and (2) CloudSat radar and CALIPSO lidar echoes are often heavily attenuated before reaching down to the liquid layer inside deep convective clouds. For every deep convection sample, P, liquid water path (LWP), and ice water path (IWP) are each averaged within ±2° about the convection center before combining into R_{CP} by Equation 1.

There is inherent difficulty in quantifying the accuracy of satellite-measured liquid water path (LWP) and ice water path (IWP) because ground truths are unavailable for these variables. The absolute values of LWP and IWP estimates are in general associated with large uncertainties. The absolute values, however, do not come into play in interpretations of the current results because relative magnitude is all that matters when R_{CP} is discussed. As such, AMSR-E LWP and CloudSat IWP offer a useful qualitative measure.

The iris effect as proposed by Lindzen et al. (2001) is built upon the assumption on changes in the precipitation efficiency of convective clouds (Lindzen & Choi, 2021, and references therein). The current choice of the CF threshold of 0.5 at all levels from 1 to 16.6 km (see above) likely favors the occurrence of deep convection, but it is worthwhile to test this presumption by running another run with non-convective (or stratiform) clouds explicitly excluded from the analysis. To this end, an ancillary composite analysis is carried out in which the same CF threshold as above is augmented with an additional constraint that all the CloudSat footprints falling within ±2° (the domain employed for computing R_{CP}) around the convection center are required to be identified as "deep convection" according to the CloudSat 2B-CLDCLASS cloud mask (Wang & Sassen, 2001).

Upper-tropospheric stability (Bony et al., 2016) is quantified as,

$$S_{\mathrm{UT}} \equiv -\frac{T}{\theta}\frac{\partial \theta}{\partial p} = \frac{RT}{c_p p} - \frac{\partial T}{\partial p} \quad \text{at 200 hPa,} \tag{2}$$

where R and c_p are the gas constant and specific heat of dry air, respectively, θ is potential temperature, T is air temperature, and p is pressure. Throughout this study, S_{UT} is defined only where AMSR-E precipitation is zero and ERA5 pressure velocity (ω) is positive so that S_{UT} is related to the radiatively driven subsidence as formulated in the stability iris hypothesis.

The study period spans 4 years from 1 January 2007 to 31 December 2010. Since AMSR-E precipitation and LWP are available only over oceans, at least 90% of CCCM pixels within ±5° (the study domain in composite space) about each convection center are required to be over ocean. Anvil clouds lingering around the convection centers may be extended to over land. The target region consists of all longitudes bound in latitude between 20°S and 20°N. This choice of latitudes is intended to ensure that the analysis is limited to the tropics by factoring out unwanted influences of the pronounced regional gradient from tropics to subtropics. Additional analyses with latitudinal boundaries raised to 30° did not essentially alter the results in any case. Indeed, the geographical distribution of the convection centers is consistently trapped within the deep Tropics regardless of S_{UT} (Figure S1 in Supporting Information S1) and of R_{CP} (Figure S2 in Supporting Information S1), implying that there is little regional bias between high and low values of S_{UT} and R_{CP}. This is because the present compositing method is conditioned on the occurrence of deep convection and, as a result, different values of S_{UT} and R_{CP} arise from variability inside the deep Tropics where convection can develop deep.

The total sample size is 3,770 (or 2,210 when screened by the deep convection mask) for the R_{CP} composite and 3,641 for the S_{UT} composite, of which roughly two thirds are from descending (1:30) tracks and the other one third from ascending (13:30) overpasses.

3. Results

3.1. Composite Cloud Fraction

The composite vertical structure of CF for different stability environments is plotted in Figure 1. The four panels in the top row represent very unstable ($S_{\mathrm{UT}} < S_1$), moderately unstable ($S_1 < S_{\mathrm{UT}} < S_2$), moderately stable ($S_2 < S_{\mathrm{UT}} < S_3$), and very stable ($S_3 < S_{\mathrm{UT}}$) upper tropospheres, where $S_1 = 27.4$ mK hPa^{-1}, $S_2 = 34.3$ mK hPa^{-1}, and $S_3 = 42.0$ mK hPa^{-1} are the first, second, and third quartiles of S_{UT}, respectively. All the four cases share a common feature that anvil clouds are spread extensively above 200 hPa from the deep convection towering at the center, except that the anvil cloud extent narrows systematically with increasing S_{UT}. The dependence on S_{UT} is clearly evidenced by Figure 1e, where the anvil CF at the level of maximum convergence is shown. Here convergence profiles were sampled from rain-free subsidence areas (i.e., conditioned on $P = 0$ and $\omega > 0$ as was for S_{UT}) and then averaged together at each level in composite space. The composite convergence profiles (Figure 1f) have a maximum near the tropopause with its magnitude diminishing with increasing S_{UT}. The interrelationship among the anvil CF, S_{UT}, and upper-tropospheric convergence derived from Figure 1 is in line with the expectations of the stability iris theory (Bony et al., 2016).

The dependence of CF on R_{CP} is shown in Figure 2. The three quartiles of R_{CP} are $R_1 = 0.619$ hr^{-1}, $R_2 = 0.942$ hr^{-1}, and $R_3 = 1.30$ hr^{-1}. CF exhibits a slight but consistent enhancement with increasing R_{CP} at levels lower than ~400 hPa as evident in Figure 2e. In contrast, upper-tropospheric CF hardly varies across the whole range of R_{CP}. It follows that an enhancement in precipitation efficiency is unlikely to have appreciable consequences on the horizontal extent of anvil clouds, although leading to a modest increase of CF in the mid and lower tropospheres.

Figure 2f shows an additional composite plot with the "deep convection" mask applied (see Section 2). It is evident that the composite CF profiles are hardly sensitive to whether or not the convection centers are explicitly screened for convective clouds. In the remainder of this paper, therefore, the results with the deep convection mask are not shown to avoid redundancy.

3.2. Cloud Radiative Effects

Next examined are the impacts of S_{UP} and R_{CP} on the CREs associated with tropical deep convection. The net CRE at the top of the atmosphere (TOA) is,

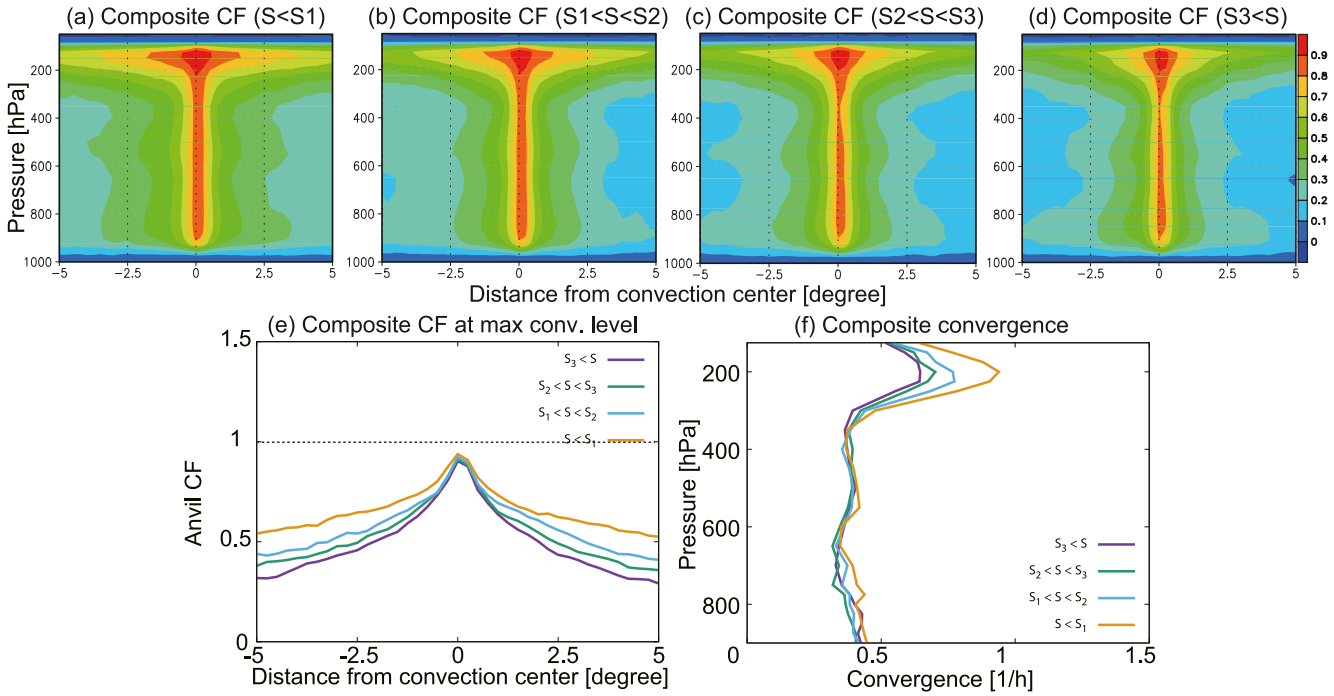

Figure 1. (a) Composite cloud fraction (CF) for $S_{UT} < S_1$, (b) CF for $S_1 < S_{UT} < S_2$, (c) CF for $S_2 < S_{UT} < S_3$, and (d) CF for $S_3 < S_{UT}$, where S_1–S_3 are the first to third quartiles of S_{UT}. (e) CF at the level of maximum convergence for different stabilities. (f) Convergence averaged horizontally at each level over rain-free subsidence areas within ±5° about the convection center.

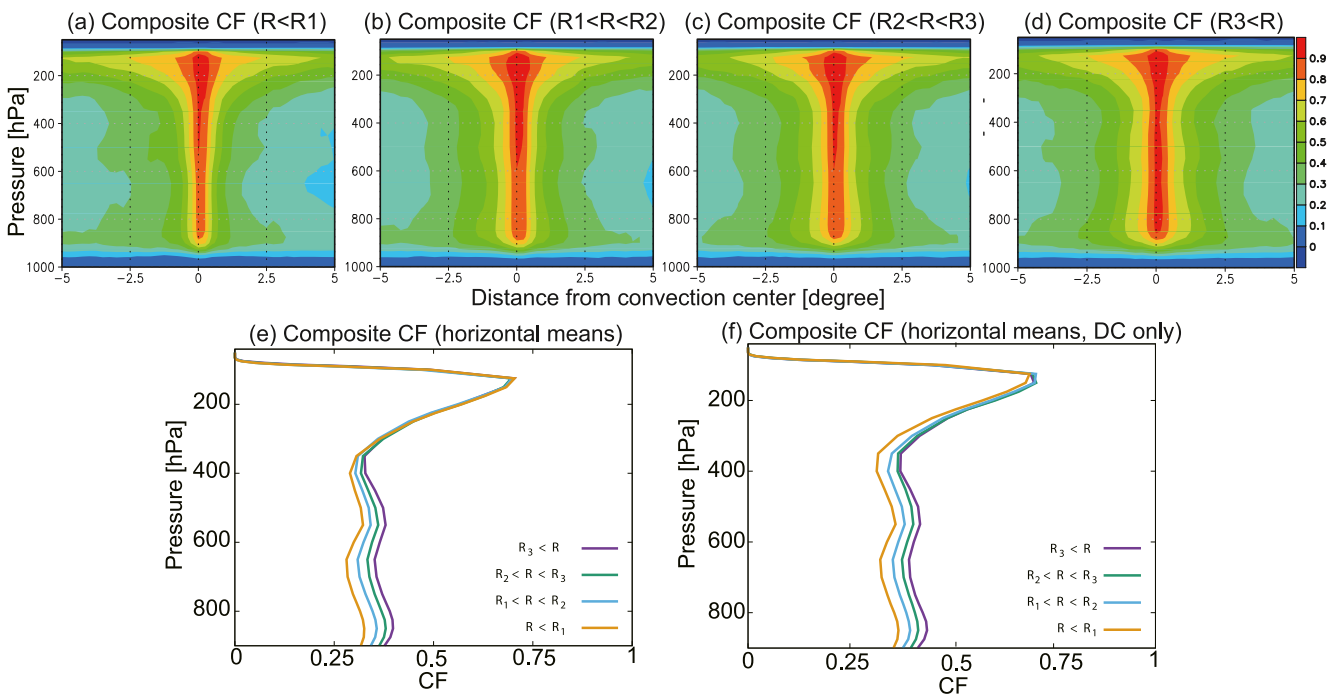

Figure 2. (a) Composite cloud fraction (CF) for $R_{CP} < R_1$, (b) CF for $R_1 < R_{CP} < R_2$, (c) CF for $R_2 < R_{CP} < R_3$, and (d) CF for $R_3 < R_{CP}$, where R_1–R_3 are the first to third quartiles of R_{CP}. (e) CF for different precipitation efficiencies, averaged horizontally at each level over ±5° about the convection center. (f) As (e) but from an ancillary analysis with the "deep convection" mask applied.

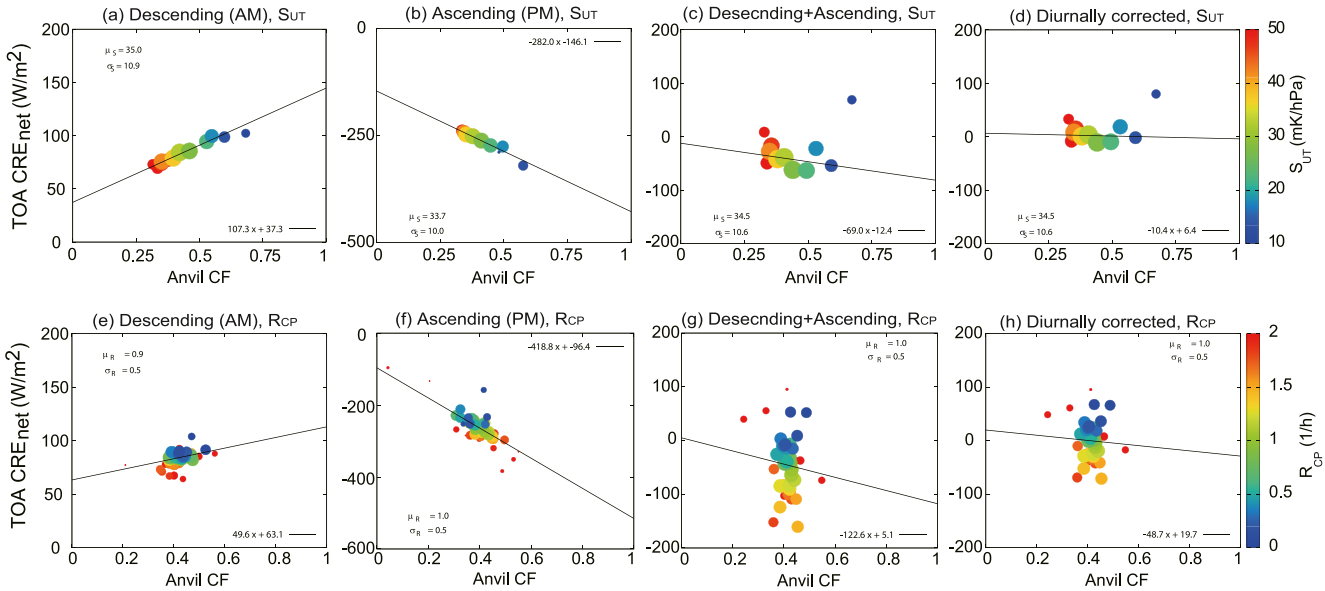

Figure 3. Top row (a–d): scatter plots between the anvil cloud fraction (CF) and the net cloud radiative effects (CRE) at top of the atmosphere sorted by S_{UT} for (a) descending tracks (midnight), (b) ascending tracks (afternoon), (c) descending and ascending tracks combined, and (d) daily mean (diurnally corrected) estimates (see text for details). Points represent different values of S_{UT} as indicated by color scale on the right. The point size is proportional to the sample size for each bin having a fixed width of $\Delta S_{UT} = 5$ mK hPa^{-1}. The mean μ, standard deviation σ, and linear regression coefficients are shown in each panel. Bottom row (e-h): As the top row but sorted by R_{CP} with a bin width of $\Delta R_{CP} = 0.05$ hr^{-1}.

$$\mathrm{CRE} = F^{\mathrm{clr}}_{\mathrm{SW}} - F^{\mathrm{all}}_{\mathrm{SW}} + F^{\mathrm{clr}}_{\mathrm{LW}} - F^{\mathrm{all}}_{\mathrm{LW}}, \tag{3}$$

where F is upwelling radiative flux at top of the atmosphere (TOA), SW and LW denote shortwave and longwave, "clr" and "all" stand for clear and all skies, respectively. Every term on the rhs is available from the CCCM product, in which the clear-sky component is a synthetic estimate computed with all clouds taken out. CRE is sorted by S_{UT} and R_{CP} as done for the composite analysis in the previous section, except that this time the convection-anvil samples are broken down into finer bins of $\Delta S_{UT} = 5$ mK hPa^{-1} and $\Delta R_{CP} = 0.05$ hr^{-1} instead of quartiles. This is for demonstration purposes: it has been confirmed that the conclusions remain unchanged with quartile bins (i.e., just four points in each scatter plot).

Figures 3a–3d show the scatter plots of the statistical relationship between the anvil CF and the net CRE at top of the atmosphere (TOA). The midnight CRE from descending satellite tracks (Figure 3a), consisting exclusively of longwave cloud warming, is tightly coupled with the anvil CF. An enhancement of CRE is accounted for by an increase of the anvil CF, which accompanies a reduction of upper-tropospheric stability (from warm to cold colors) as we have seen in Figure 1. Afternoon observations of the net CRE (Figure 3b), by contrast, imply a striking shortwave effect largely overwhelming the longwave effect. When the morning and afternoon measurements are averaged together, the resulting CRE tends to fall on the negative side (i.e., have a potential of cooling the planet) with its magnitude weakly increasing with the anvil CF (Figure 3c). Note that since the size of deep convection samples is larger for midnight (descending) tracks than for afternoon (ascending) tracks, Figure 3b would be even more dominated by shortwave cooling if the ascending and descending tracks are individually averaged first and then combined together. This day-night combined CRE, however, is not an adequate measure of the daily mean CRE because the daytime equatorial crossing time of 13:30 is nearly the peak hour of incoming solar radiation. To remedy this, a simplistic diurnal correction is applied assuming a semi-sinusoidal diurnal cycle of solar irradiation,

$$\overline{F}_{\mathrm{SW}} = \frac{\pi I_0}{12} \int_6^{18} \left[-\cos\left(\frac{\pi t}{12}\right) \right] dt = 2I_0, \tag{4}$$

where $\overline{F}_{\mathrm{SW}}$ is the daily mean shortwave flux and t is local time in hour (e.g., Shinoda & Hendon, 1998). The amplitude πI_0 is eliminated with a given A-Train measurement at $t = 13.5$ hr or $F^{\mathrm{obs}}_{\mathrm{SW}} = -\pi I_0 \cos(13.5\pi/12)$ as

$$\overline{F}_{\mathrm{SW}} = -\frac{2F_{\mathrm{SW}}^{\mathrm{obs}}}{\pi \cos(13.5\pi/12)} \approx 0.638 F_{\mathrm{SW}}^{\mathrm{obs}}. \quad (5)$$

Equation 5 relies on a number of naive assumptions (e.g., cloud properties do not vary with local time) and is only intended to offer an approximate (yet useful) correction factor. The diurnal variation of cloud properties (e.g., Chepfer et al., 2019) may modify the corrected CRE to a certain extent. Figure 3d shows the daily mean net CRE with the above correction applied. Interestingly, the diurnally corrected shortwave CRE almost precisely offsets the longwave CRE for almost all anvil CFs.

The scatter plots broken down by R_{CP} are presented in Figures 3e–3h. The correlation of the anvil CF with the net CRE is recognizable when midnight and afternoon observations are separated, whereas no longer visually discernible in the daily mean (Figure 3h). Unlike the S_{UT} cases, color gradient is primarily vertical across the line of zero daily-mean CRE, implying that a low (high) R_{CP} is likely to accompany a positive (negative) net CRE for reasons unrelated to anvil clouds. This correlation is presumably ascribed to variability in the mid-to low-level CFs as was found in Figure 2e.

4. Conclusion and Discussion

A-Train satellite observations and ERA5 datasets were analyzed to investigate the physical processes considered to be crucial for the iris effect, namely the cloud-water to precipitation ratio R_{CP} (a substitute for precipitation efficiency) and upper-tropospheric stability S_{UT}. Composite analysis was carried out to illustrate the spatial structure of deep convective clouds and the accompanying anvil clouds, and to examine how CREs vary with the anvil CF. The composite samples are separated by S_{UT} and R_{CP} in search of evidence for the possible roles of these parameters behind the iris effect. The present analysis is focused on anvil clouds accompanying rather extreme convection reaching higher than 16 km. The cloud-top height of deep convection was chosen in this way to facilitate comparison with upper-tropospheric stability at the fixed level of 200 hPa. Anvil clouds formed at somewhat lower levels will be investigated in future work.

A main conclusion is, in line with a previous study (Saint-Lu et al., 2020), that S_{UT} is clearly linked with the anvil CF as hypothesized by the stability iris theory. To the contrary, the anvil CF is invariant over the whole range of R_{CP}, although the mid-to low-level CFs modestly but systematically increase with R_{CP}. A possible explanation for this coherent variability in mid- and low-level CFs may be given in light of the mesoscale organization of convective systems. Sui et al. (2020) argued that precipitation efficiency rises as the degree of convective organization increases (i.e., the stratiform rain fraction becomes higher). Choi et al. (2017) underscored the key roles of stratiform rain clouds in the context of the iris effect. Changes in the mid-to low-level CFs with R_{CP} are presumably relevant to the observed modulation in CREs, although in a way at odds with what Lindzen et al. (2001) hypothesized with precipitation efficiency.

Choi et al. (2017) found that a higher precipitation efficiency (the ratio of convective rain to the total rain in their definition) is associated with a lower cirrus fraction over the tropical west Pacific, not necessarily in line with the present findings. The disagreement is presumably attributable to multiple reasons, including the differences in the definition of precipitation efficiency and in analysis strategies. The analysis of Choi et al. (2017) was based on monthly-mean regional statistics of satellite observations, while the current conclusions are derived from the composite statistics constructed with event-by-event CF profiles. Another possible source of uncertainty lies in the definition of convective clouds. The Tropical Rainfall Measuring Mission Precipitation Radar data were adopted by Choi et al. (2017) to separate convective and stratiform rains, while the CloudSat cloud mask is used in this study.

The current work does not seek evidence for the climatic consequences of the processes investigated here, but nevertheless it would be beneficial to briefly discuss the implications for cloud feedback. The longwave CRE at midnight infers a negative feedback just as predicted by the stability iris hypothesis, while the noontime CRE suggests a weak positive feedback in which the shortwave heating owing to a shrinkage of anvil clouds slightly outruns the longwave effect. These competing effects are estimated to largely cancel each other out when averaged over a diurnal cycle. The stability iris effect is suggested to be nearly radiatively neutral in a climatological context, although away from neutrality on subdaily time scales.

The shortwave and longwave CREs of tropical cirrus have been long known to have a tendency of offsetting each other (e.g., Hartmann et al., 2001; Ramanathan et al., 1989). The near cancellation of CRE is rarely achieved on a cloud-by-cloud basis but results from an ensemble of clouds with different altitudes and optical depths averaged together (Hartmann & Berry, 2017). The dynamical, microphysical, and radiative properties of anvil clouds are intertwined with the convective life cycle (Gasparini et al., 2019; Masunaga & Bony, 2018; Takahashi et al., 2017), and so would be the transient CRE imbalances on a subdaily time scale. These short-term processes remain a missing piece of puzzle to draw the whole picture of the anvil cloud feedback.

Data Availability Statement

The CCCM Release D1 product is available from https://doi.org/10.5067/AQUA/CERES/CCCM-FM3-MODIS-CAL-CS_L2.RELD1 (NASA/LARC/SD/ASDC, 2011). The Remote Sensing Systems Advanced Microwave Scanning Radiometer for the Earth Observing System data may be downloaded from https://www.remss.com/missions/amsr/. The European Centre for Medium-Range Weather Forecasts Reanalysis version 5 data set is available at https://doi.org/10.24381/cds.bd0915c6 (Hersbach et al., 2018).

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Erratum

In the originally published version of this article, delineated author contributions were included accidentally. This contribution list has been removed, and the present version may be considered the authoritative version of record.