1 2	Observational evidence for a stability Iris effect in the Tropics
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6	Key Points:
7	• Spaceborne lidar observations show that anvil clouds rise and reduce their cov-
8	erage when the tropics warm
9	• Observations and meteorological reanalyses support the stability-Iris effect mech-
10	anism
11	• There is evidence for a stability Iris effect over a large range of spatial scales

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#### 12 Abstract

Anvil clouds cover extensive areas of the tropics, and their response to global warming 13 can affect cloud feedbacks and climate sensitivity. A growing number of models and the-14 ories suggest that, when the tropical atmosphere warms, anvil clouds rise and their cov-15 erage decreases, but observational support for this behavior remains limited. Here we 16 use 10 years of measurements from the spaceborne CALIPSO lidar to analyze the ver-17 tical distribution of clouds and isolate the behavior of anvil clouds. On the interannual 18 time-scale, we find a strong evidence for anvils rise and coverage decrease in response 19 to tropical warming. Using meteorological reanalyses, we show that this is associated with 20 an increase in static stability and with a reduction in clear-sky radiatively-driven mass 21 convergence at the anvils height. These relationships hold over a large range of spatial 22 scales. This is consistent with the stability Iris mechanism suggested by theory and mod-23 eling studies. 24

# 25 Plain Language Summary

Anvil clouds cover about 40% of the tropics. Their response to global warming, es-26 pecially changes in their height or in their horizontal extent, has the potential to affect 27 the Earth's surface temperature. By analyzing 10 years of observations of the vertical 28 distribution of clouds from a spaceborne lidar, we show that the anvils rise and reduce 29 their coverage during the years that are anomalously warm. By using meteorological re-30 31 analyses, we further show that this behavior is consistent with the stability Iris effect suggested by theory and modeling studies. These results improve our physical understand-32 ing of the response of tropical clouds to warming, and present relationships that may be 33 used to test climate models. 34

# 35 1 Introduction

Anyil clouds cover extensive areas of the tropics, reflect solar radiation and reduce 36 outgoing long-wave radiation. In case global warming would affect their height and cov-37 erage, this could influence climate sensitivity (Zelinka & Hartmann, 2010; Hartmann, 38 2016; Su et al., 2017). The Fixed Anvil Temperature (FAT) hypothesis states that anvils 39 rise nearly isothermally when the tropics warm (Hartmann & Larson, 2002). Accord-40 ing to FAT, anvils are formed by convective detrainment, and the altitude of maximum 41 detrainment is constrained, through mass conservation, by the convergence of mass in 42 the surrounding clear-sky upper troposphere. The latter is due to the vertical gradient 43 of subsidence, which is primarily driven by the decrease with height of the radiative cool-44 ing by water vapor. The variation of water vapor with height is very much constrained 45 by the Clausius-Clapeyron thermodynamical relationship, and therefore the tempera-46 ture at which the clear-sky radiative cooling drops is relatively invariant with surface tem-47 perature. When the surface warms, it rises in step with the isotherms, and therefore the 48 peaks of the clear-sky radiatively-driven mass convergence and of the associated convec-49 tive detrainment rise. Zelinka and Hartmann (2010) refined this theory and formulated 50 the Proportionally Higher Anvil Temperature (PHAT) hypothesis, which states that anvils 51 do not rise strictly isothermally but slightly warm instead, due to the sharp increase of 52 static stability with height in the upper troposphere. Several studies have provided ob-53 servational support for this theory, showing the rise of high clouds during the very strong 54 1997-98 El Niño event (Xu et al., 2005), the invariance of high clouds temperature rel-55 ative to surface temperature over a 6-months period (Xu et al., 2007; Eitzen et al., 2009), 56 and the good correspondence between vertical profiles of cloud fraction and radiatively-57 driven clear-sky mass convergence over a 10-months period (Kubar et al., 2007). These 58 observations were then confirmed on longer periods of 4 to 6 years (Zelinka & Hartmann, 59 2011; Li et al., 2012; Thompson et al., 2017). 60

Analyzing geostationary satellite data, Lindzen et al. (2001) suggested that anvils 61 also reduce their coverage when the tropics warms. This behavior was referred to as an 62 "Iris effect", by analogy with the eye's iris. These observational results have been rebut-63 ted, primarily on methodological grounds (Hartmann & Michelsen, 2002; Del Genio & Kovari, 2002). Nevertheless, whether or not an Iris effect operates in climate remains an 65 open question (Bony et al., 2015; Mauritsen & Stevens, 2015). Zelinka and Hartmann 66 (2011) observed a reduction in the tropical high clouds cover when the tropics warm, us-67 ing various satellite data including CloudSat radar observations (Stephens et al., 2017), 68 which were only available for 4 years at that time (including one El Niño event and one 69 La Niña event). Choi et al. (2017) focused on the western tropical Pacific, and observed 70 that convective clouds "concentrate" when the sea surface temperature rises. However, 71 observing such a limited region does not allow to conclude on inherent cloud responses 72 to surface temperature changes, since cloud systems can shift in and out of the box, due 73 to dynamical effects. More recently, using various satellite observations over 13 years for 74 the longest, Su et al. (2017) and Liu et al. (2017) reported a decrease of tropical high 75 cloud fraction in response to interannual surface warming. Su et al. (2017) proposed that 76 it was linked to the tightening of the ascending branch of the Hadley circulation, although 77 the mechanism underlying the cloud fraction decrease was not investigated in details. 78

In models, tropical warming can also be associated with a reduction of anvils cov-79 erage (Tompkins & Craig, 1999; Zelinka & Hartmann, 2010; Khairoutdinov & Emanuel, 80 2013; Bony et al., 2016; Su et al., 2017; Cronin & Wing, 2017). With the idea that the 81 reduction of anvils coverage could be linked to their elevation and to PHAT, Bony et al. 82 (2016) proposed a thermodynamic "stability Iris" hypothesis. As the tropics warm, the 83 anvils rise and find themselves in a more stable atmosphere, due to the lower air pressure that increases static stability. The stability Iris hypothesis states that the increased 85 stability reduces the magnitude of the radiatively-driven clear-sky mass convergence at 86 the height of anvil clouds, thus weakening convective detrainment at that height, lead-87 ing to a reduction of the anvils coverage. Several climate models and convection-resolving 88 models have provided support for this hypothesis (Zelinka & Hartmann, 2010; Bony et 89 al., 2016; Cronin & Wing, 2017), but the existence of the stability Iris effect in Nature 90 remains an open question. 91

To investigate the existence of the stability Iris effect in observations, we use lidar 92 observations derived from the A-train Cloud-Aerosol Lidar and Infrared Pathfinder Satel-93 lite Observations CALIPSO (Stephens et al., 2017; Winker et al., 2017), together with 94 ERA5 meteorological reanalyses (Hersbach et al., 2019). Lidar measurements from CALIPSO 95 provide the most accurate cloud vertical profile measurements without time drift (Winker 96 et al., 2007, 2017) and on a very fine vertical resolution (60 m). Here we use 10 years 97 of data, that include two El Niño and three La Niña events, to examine the behavior of 98 the anvil cloud fraction as the tropics undergo warming or cooling events. In order to qq test the robustness of our results, and to determine whether they could emerge at coarser 100 vertical resolutions of climate models, we also use the GCM-oriented CALIPSO Cloud 101 Product GOCCP (Chepfer et al., 2010). We further investigate the existence of PHAT 102 in these observations, as well as the spatial scale at which PHAT and the stability Iris 103 potentially hold. 104

#### 2 Data and methods 105

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# 2.1 Detection of anvils in CALIPSO observations

In this study, we refer to anvils as the detraining top of deep convective clouds, which 107 lie above 8 km in the tropics (Yuan et al., 2011). To identify tropical anvils, we use the 108 monthly-mean three dimensional cloud fraction derived from the lidar level 3 cloud oc-109 currence product (CAL\_LID\_L3\_Cloud\_Occurrence-Standard-V1-00, Winker (2018)), here-110 after CALIPSO-Cloud-Occurrence. CALIPSO-Cloud-Occurrence is gridded on 2° lat-111

itude  $\times 2.5^{\circ}$  longitude and on the native CALIOP vertical resolution of 60 m (344 ver-112 tical levels). We use it over the tropical belt (30N-30S), averaged over day and night, 113 from June 2006 to December 2016 (10 years). CALIPSO-Cloud-Occurrence is used through-114 out the study except for what is related to Figure 4b in section 4.1, where results are repli-115 cated using GOCCP instead (Chepfer et al., 2010), which provides the monthly-mean 116 cloud fraction gridded on a  $2^{\circ} \times 2^{\circ}$  horizontal grid with a degraded vertical resolution of 117 480 m (40 vertical levels), from June 2006 to December 2017 (11 years). In this case, GOCCP 118 is used to determine whether our results can emerge at coarser vertical resolutions of Gen-119 eral Circulation Models (GCMs), indicating whether the ability of climate models to re-120 produce these results can be tested. 121

For each month and location  $(2^{\circ} \times 2.5^{\circ} \text{ gridbox})$ , we define the altitude and the cloud 122 fraction of anvil clouds, resp.  $Z_{anv}$  and  $CF_{anv}$ , where there is a local maximum of op-123 tically thick but non-opaque ice clouds coverage (optical depth  $0.3 \leq \tau < 5$ ), impos-124 ing that this maximum occurs above 8 km and that its cloud fraction exceeds a thresh-125 old value referred to as  $CF_c$  (we use  $CF_c = 0.03$ ). If there are several local maxima, 126 we consider the closest to the cloud fraction centroid. Locations that are cloud-free or 127 where no maximum exceeds  $CF_c$  are masked and ignored when tropically-averaging  $Z_{anv}$ 128 or  $CF_{anv}$ . With CALIPSO-Cloud-Occurrence only, local maxima are identified from the 129 smoothed cloud fraction profile, using a 480-meters running window (smoothing is only 130 used for this purpose). Note that our results do not critically depend on the method used 131 to locate the anvils altitude (not shown). 132

This definition of anvils excludes clear-sky regions and regions with a low cloud frac-133 tion, as well as subvisible cirrus clouds, and the cores of deep convective clouds. These 134 choices will be tested and discussed in section 4. As will be shown, the conclusions of 135 this study do not critically depend on the value of  $CF_c$ , nor on the exact range of op-136 tical depths considered. Figure 1a shows the annual-mean cloud fraction profile averaged 137 over locations where anyils have been detected (locations void of anyils are ignored), within 138 the tropical belt (30N-30S), considering thick but non-opaque ice clouds with  $CF_{anv} \geq$ 139 0.03.140



Figure 1. Vertical profiles, averaged over the tropics and over 2006-2016, of: a) cloud fraction averaged over locations with anvils only (blue) and radiatively-driven clear-sky mass convergence  $(\partial \omega_r / \partial p, \text{ orange})$ , b) atmospheric static stability (S, red) and temperature (grey), c) clear-sky radiative cooling ( $Q_r$ , purple) and radiatively-driven clear-sky pressure velocity ( $\omega_r$ , green).

#### 2.2 Clear-sky variables from ERA5

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According to PHAT, anvils form at the altitude of the peak of upper-tropospheric horizontal mass divergence in convective regions, which coincides with a peak of radiativelydriven mass convergence in clear-sky regions (Hartmann & Larson, 2002; Zelinka & Hartmann, 2010). The radiatively-driven clear-sky pressure velocity ( $\omega_r$ , positive downward) and divergence ( $D_r$ ) are diagnosed as:

$$D_r = \max\left(\frac{\partial\omega_r}{\partial p}\right)$$
, with  $\omega_r = \frac{Q_r}{S}$  and  $S = \frac{T}{p}\frac{R}{c_p} - \frac{\partial T}{\partial p}$  (1)

where  $Q_r$  is the clear-sky radiative cooling rate, p the air pressure, S the static stability, T the temperature, R the gas constant,  $c_p$  the isobaric specific heat of dry air.

We use monthly mean clear-sky radiative cooling, temperature and pressure from ERA5 reanalyses in the tropics, horizontally regridded onto the CALIPSO-Cloud-Occurrence  $2^{\circ} \times 2.5^{\circ}$  grid, with a vertical resolution of about 300 m in the upper troposphere (137 vertical levels in the whole atmosphere), from June 2006 to December 2017. We compute the peak of upper-tropospheric horizontal mass divergence at each month and location, which is simply defined as its maximum. The altitude of this peak is referred to as  $Z_{Dr}$ .

Figures 1b and 1c show that the annual-mean tropically-averaged  $Q_r$  (in purple) 155 weakens in the upper-troposphere above 10 km, while S (in red) sharply increases above 156 12 km. As a consequence, the radiatively-driven clear-sky subsidence ( $\omega_r$ , in green) weak-157 ens above 10 km. Following this, Figure 1a shows that the radiatively-driven clear-sky 158 mass convergence  $(\partial \omega_r / \partial p)$ , in orange) peaks near 12 km, at about 216 K (Fig. 1b), and 159 the anvils cloud fraction profile (in blue) peaks near 12.5 km, at about 214 K. The anvils 160 cloud fraction profile thus exhibits a clear correspondence with the radiatively-driven clear-161 sky mass convergence profile. 162

## 2.3 Surface temperatures from HadCRUT4

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Monthly land and sea surface temperatures  $(T_s)$  are derived from HadCRUT4 (Morice et al., 2012) in the tropics, from June 2006 to December 2017. The Oceanic Niño Index (ONI, Golden Gate Weather Services) indicates that the 11-year period, or the 10-year period from 2006 to 2016, includes one strong (2009-2010) and one very strong (2015-2016) El Niño events, two strong (2007-2008, 2010-2011) and one moderate (2011-2012) La Niña events. El Niño years correspond to the highest tropical-mean surface temperature and La Niña years to the lowest.

# <sup>171</sup> 3 Evidence for PHAT and the stability Iris effect

<sup>172</sup> We consider yearly means, computed by averaging each year from July to June, <sup>173</sup> in order to capture the response to El Niño/La Niña-induced surface temperature changes, <sup>174</sup> that will be maximum during boreal winter (we compute tropically-averaged yearly anoma-<sup>175</sup> lies as  $\langle \overline{x} \rangle_{year} - \langle \overline{x} \rangle_{2006-2016}$ , where the overbar denotes tropical averaging, brackets de-<sup>176</sup> note time averaging, *year* refers to each year within 2006-2016, and *x* can be  $Z_{anv}$ ,  $T_S$ , <sup>177</sup>  $Z_{D_r}$ , or other variables).

On the tropical and interannual scales, strong correlations of  $Z_{anv}$  and  $Z_{D_r}$  with 178 the tropical-mean surface temperature confirm that both anvils and  $D_r$  rise as the trop-179 ics warm (r = 0.87 and r = 0.94, respectively, Figures 2a and 2b). The altitudes of 180 both anvils and  $D_r$  are the highest during the very strong 2015-2016 El Niño, relative 181 to the rest of the 10-year record, with anvils forming about 100 m higher than on av-182 erage (Figure 2a). During that very strong El Niño year, the air temperature at around 183 12 km increased by about 0.84 K relative to the 10-year average, while the anvils tem-184 perature remained at approximately the same temperature, changing by only 0.006 K 185 (not shown). The altitude of anvils varies in phase with the altitude of the divergence 186 peak  $D_r$ , as shown by the correlation of r = 0.95 between  $Z_{anv}$  and  $Z_{D_r}$ , with a slope 187 that is not statistically different from 1 (Figure 2c). All these findings support the PHAT 188

# hypothesis. Note that models rather predict that anvils migrate more than the diver-

<sup>190</sup> gence peak (Zelinka & Hartmann, 2010).



Figure 2. Scatterplots showing tropically-averaged yearly anomalies (each dot is a July-to-June year, anomalous relative to the time-mean over 2006-2016). Linear regressions are also reported (black lines) together with the Pearson correlation coefficient r, the p-value and the slope. Colors show the tropical-mean surface temperature anomaly and numbers in (a) give the corresponding year of El Niño (red) and La Niña (blue) events. The relationships involve the anvils altitude  $Z_{anv}$ , the tropical surface temperature  $T_S$  and the divergence peaks altitude  $Z_{D_r}$ .

In addition to PHAT, Figure 3a shows that tropical warming is associated with a reduced anvils coverage (negative correlation between  $CF_{anv}$  and  $T_s$ , r = -0.97). This supports the existence of an Iris effect, as defined by a reduction of anvils coverage associated with tropical warming.

<sup>195</sup> Moreover, the decrease of anvils coverage is associated with a weakening of the ra-<sup>196</sup> diative divergence peak  $D_r$ , as shown by the correlation between  $CF_{anv}$  and  $D_r$  of r =<sup>197</sup> 0.84 (Figure 3b). This supports the idea that anvils are tightly linked to the peak of the <sup>198</sup> radiatively-driven clear-sky convergence: not only their altitudes but also their ampli-<sup>199</sup> tudes vary in phase.

As the clear-sky radiative cooling profile  $(Q_r)$  shifts upward, the vertical gradient 200 of the radiatively-driven subsidence  $(\partial \omega_r / \partial p)$  becomes less steep because of the stronger 201 stability (S), which weakens the divergence peak  $D_r$  (Eq. 1). This is supported by the 202 negative correlation between  $D_r$  and the stability at the level of  $D_r$   $(S_{D_r})$ , of r = -0.82203 (Figure 3c). This all happens in response to tropical warming, as shown by the corre-204 lation between  $S_{D_r}$  and  $T_s$  of r = 0.89 (Figure 3d). To summarize, the reduction of both 205  $D_r$  and  $CF_{anv}$  is associated with an increase of static stability at the height of anvils, 206 in response to tropical warming. Although causality cannot be determined from obser-207 vations, these relationships are fully consistent with the stability Iris mechanism. 208

<sup>209</sup> 4 Spatial scale and robustness of the observed relationships

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# 4.1 Varying spatial scale, critical anvil cloud fraction and vertical resolution

To test the robustness of our results, we assess the sensitivity of the PHAT relationships shown on Figures 2a and 2c ( $Z_{anv}$  against  $T_S$  and against  $Z_{D_r}$ ), the Iris effect and the stability Iris effect relationships shown on Figures 3a and 3b ( $CF_{anv}$  against  $T_S$ 



Figure 3. As Figure 2, but for other relationships involving the anvils cloud fraction  $CF_{anv}$ , the divergence peak  $D_r$ , the tropical-mean surface temperature  $T_S$  and the stability at  $D_r$  level  $(S_{D_r})$ .

and against  $D_r$ ), to the value of the  $CF_c$  threshold used to identify anvils. Figure 4a shows that all relationships remain strong ( $|r| \ge 0.7$ ) and significant (*p*-value < 0.05) for  $CF_c$ ranging from 0 to 0.13.

PHAT and the stability Iris hypothesis relate the altitude and coverage of anvils to the radiatively-driven convergence in surrounding clear-sky regions. However, the scale of this "surrounding" is unclear. Thompson et al. (2017) observed that PHAT holds on zonal averages at each latitude. This leads to the following question: over which spatial scale does the balance between clear-sky convergence and convective detrainment hold?

To answer this question, we use the different values of  $CF_c$  tested above, because 223 they affect the tropical coverage of identified anvils. For example,  $CF_c = 0$  (i.e.  $CF_{anv} >$ 224 0) means that any large defined where there is a cloud fraction maximum in the upper 225 troposphere without any constraint on the magnitude of this maximum (in this case anvils 226 are present in about 83% of the  $2^{\circ} \times 2.5^{\circ}$  grid-points of the tropics on average), while  $CF_c =$ 227 0.13 (i.e.  $CF_{anv} \ge 0.13$ ) means that anvils are only detected where the cloud fraction 228 maximum in the upper troposphere exceeds 0.13 (represents about 30% of the  $2^{\circ} \times 2.5^{\circ}$ 229 grid-points of the tropics on average, Figure 4a). On the other hand, the clear-sky re-230 gions (hence  $D_r$  and  $Z_{D_r}$ ) are so far considered over the whole tropical belt in all cases 231 (30N-30S). The PHAT, Iris and stability Iris relationships are thus already shown to be 232 well observed at the tropics scale, since they remain strong even when some clear-sky 233 regions are remote from the identified anvils (for all values of  $CF_c$ ). 234



Figure 4. Evolution of linear regressions shown on Figures 2a, 2c, 3a and 3b with varying the spatial scale and critical cloud fraction  $CF_c$ , with Calipso-Cloud-Occurrence (a) and GOCCP (b), and with varying optical depths of selected anvils (c). Color bars give the absolute value of the Pearson correlation coefficients r. Hatched light bars indicate that p-value  $\geq 0.05$ . The grey line gives the average percentage of the tropics covered by identified anvils. a, b) The last case (Anvil scale) is when the definition of  $T_S$ ,  $D_r$  and  $Z_{D_r}$  is restricted to locations with anvils. c) Only clouds of a certain range of optical depths ( $\tau$ ) are considered, above 8 km: all clouds (ice and water), ice opaque clouds, and ice non-opaque clouds within different  $\tau$  ranges. The dashed blue line gives the average altitude of selected anvils  $Z_{anv}$  in annual-mean. The dotted brown horizontal line indicates the annual-mean value of  $Z_{D_r}$ , averaged over the whole tropics. a, c) Categories in bold correspond to the results shown in section 3.

We now restrict the definition of the clear-sky surroundings to the immediate vicin-235 ity of anyils which exceed a strong cloud fraction threshold, by computing  $D_r$  and  $Z_{D_r}$ 236 only on locations where anvils have been detected with  $CF_{anv} \ge 0.13$  (the surface tem-237 perature is also considered only on these same locations). In this case, we choose the strongest 238  $CF_c$  threshold because it corresponds to the smallest scale we can investigate, as most 239 clear-sky regions are then remote from the identified anvils. These remote clear-sky re-240 gions are thus fully ignored and relationships can be examined at the anvil scale (or lo-241 cal scale). The corresponding correlation coefficients are shown on Figure 4a at the ex-242 treme right of the plot, and remain strong and significant for all four relationships. The 243 same results are obtained when using the total cloud fraction (that is, not restricting to 244 ice thick non-opaque clouds), with all correlations remaining strong and significant for 245 the whole range of spatial scales (not shown). The PHAT, Iris and stability Iris relation-246 ships thus hold both at the large scale, such as those of the Hadley-Walker circulations, 247 and at the scale of the close surroundings of the anvils, within a few hundred kilometers. 248

We now replicate this analysis with the GCM-oriented GOCCP product, which de-249 tects and diagnoses the total cloud fraction assuming a coarser vertical resolution of the 250 lidar backscatter signal, of 480 m instead of 60 m. Here again, depending on the  $CF_c$ 251 threshold used to define anvils, anvils cover either 99% (with  $CF_{anv} > 0$ ), or about 70% 252 (with  $CF_{anv} \ge 0.07$ ), or about 30% (with  $CF_{anv} \ge 0.23$ ) of the 2°×2° grid-points of 253 the tropics on average (Figure 4b). Although all four correlations weaken with GOCCP, 254 they are still significant when anvils cover 99%, 70% or 30% of the tropics (|r| > 0.6). 255 Only at the anvil scale, where clear-sky regions are restricted to the immediate vicin-256 ity of anvils selected with a strong threshold ( $CF_{anv} \geq 0.23$ ), evidence of the stabil-257 ity Iris effect becomes faint but is still detectable (r = 0.52 with p-value = 0.0992). There-258 fore the PHAT, Iris and stability Iris relationships remain detectable with a degraded 259 vertical resolution of about 500 m, although with better significance when clear-sky re-260 gions encompass more than the immediate vicinity of anvils. 261

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#### 4.2 Influence of the anvils optical depth

So far, our definition of anvil clouds has been restricted to thick but non-opaque 263 ice clouds  $(0.3 \le \tau < 5)$ . Figure 4c compares the correlation coefficients for the PHAT, 264 Iris and stability Iris relationships when using a range of optical depths. Note that the 265 value of  $CF_c$  is adapted in each case to remain proportional to the maximum of the annual-266 mean tropically-averaged cloud fraction profile; practically  $CF_c = 0.42 \times \max(\langle \overline{CF} \rangle)$ . 267 As the optical depth decreases, the altitude of the selected clouds increases, going up to 268 15 km for thin ice clouds ( $\tau < 0.01$ ), consistent with the persistent occurrence of sub-269 visible cirrus clouds near the tropical tropopause (e.g. Wang et al. (1994); Jensen et al. 270 (1996, 1999)). On the other hand, the altitude of opaque ice clouds, which correspond 271 272 to the cores of deep convective clouds, is around 12 km.

The PHAT relationships remain strong and significant for all optical depths (r > 0.7), meaning that the altitude of all high clouds (opaque, thick or thin) is correlated with  $T_S$  and  $Z_{D_r}$ . PHAT thus seems to hold for anvils as well as for high cirrus clouds, which is consistent with the vertical structure of the atmosphere, including cirrus clouds near the tropopause, rising approximately in step with the atmospheric isotherms as the tropics warm (Gage & Reid, 1986; Lu et al., 2008).

The Iris and stability Iris relationships remain strong and significant for most nonopaque ice clouds (|r| > 0.7 for  $\tau \ge 0.03$ ), except for high subvisible cirrus clouds (|r| < 0.5 for  $\tau < 0.01$ ). This suggests that the cloud fraction of subvisible clouds is dominated by in-situ cirrus clouds, which form when ice condensates near the tropopause rather than being directly injected from deep convection, regardless of the stability Iris effect. The Iris and stability Iris relationships are also weak and non-significant for ice opaque clouds (0.5 < |r| < 0.6 for Iris and r < 0.4 for stability Iris), which are found at the deepest cores of convective systems (as suggested by their relatively low altitude, tropical coverage and strong optical depth). This suggests that the cloud fraction of deep convective cores is less constrained by the upper-tropospheric clear-sky mass divergence than the cloud fraction of detrained anvils. Following this, the ratio of  $CF_{anv}$  for opaque ice clouds to  $CF_{anv}$  for thick but non-opaque ice clouds  $(0.3 \le \tau < 5)$  is the greatest (0.88) during the hottest 2015-2016 El Niño year, and the second lowest (0.83) during the coldest 2007-2008 La Niña year, although there is no correlation with  $T_S$  on the 10-year period (not shown).

#### <sup>294</sup> 5 Conclusions

Interannual variations of the anvils cloud fraction, inferred from 10 years of CALIPSO 295 measurements, show that the anvils coverage is reduced when the tropics are anomalously 296 warm. ERA5 reanalyses further show that the altitude and extent of anvils vary in phase 297 with the altitude and strength of the radiatively-driven clear-sky mass convergence peak, 298 and that both are tightly linked to the static stability profile. As the tropics warm, the 299 peak of the clear-sky radiatively-driven mass convergence rises, and weakens because of 300 the increase in stability with height, resulting in a reduced anvil cloud fraction. High sub-301 visible cirrus clouds and deep-convective clouds also rise in step with temperature, but 302 only the cloud fraction of anyil clouds is reduced with warming, consistent with the sta-303 bility Iris effect. Robust, consistent and highly significant relationships derived from 10 years 304 of CALIPSO observations provide a strong observational support for both the PHAT and 305 the stability Iris hypotheses. Observations further show that PHAT and the stability Iris 306 effect hold over a large range of spatial scales. This suggests that clear-sky regions can 307 influence anvils both in the vicinity of clouds and remotely over long distances. 308

At the interannual scale, tropical warming anomalies generally coincide with El Niño 309 events. Although the interannual and the longer-term climate change responses to trop-310 ical warming cannot be directly compared, theoretically PHAT and the stability Iris ef-311 fect can hold in both contexts. Whether or not the relationship with temperature shown 312 here actually applies to longer-term global warming, remains an open question. Beyond 313 surface warming, anthropogenically-induced changes in upper-tropospheric  $CO_2$  and ozone 314 concentrations might change the static stability profile, and thus potentially overwhelm 315 the stability Iris effect (Harrop & Hartmann, 2012). CO<sub>2</sub>-induced changes in the atmo-316 spheric overturning circulation (Bony et al., 2013) could also influence the altitude and 317 coverage of tropical anvils. The continuation of spaceborne lidar measurements on the 318 long term will allow us to monitor these changes. Although the evidence for the stabil-319 ity Iris effect is stronger when using highly vertically-resolved lidar measurements, it re-320 mains when using data at coarser vertical resolutions. This suggests that GOCCP could 321 be used to test the ability of climate models to reproduce this effect. This will consti-322 tute a necessary, albeit not sufficient, test of the credibility of the predicted behavior of 323 anvil clouds with temperature in the models. 324

Finally, we emphasize that the stability Iris hypothesis does not imply anything about the radiative impact of the anvils behavior. The rise of anvils with warming is known to produce a positive climate feedback (Zelinka & Hartmann, 2010). The decrease of anvils coverage with warming can be associated with both an increase in the outgoing longwave radiation and a decrease in planetary albedo. It can also enhance the exposure to space of low-level clouds, which may also impact the overall planetary albedo. Whether one effect dominates over the other is unknown a priori, and will require a specific study.

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# 347 **References**

348	Bony, S., Bellon, G., Klocke, D., Sherwood, S., Fermepin, S., & Denvil, S. (2013,
349	June). Robust direct effect of carbon dioxide on tropical circulation and re-
350	gional precipitation. Nature Geoscience, $6(6)$ , 447–451. Retrieved 2013-09-30,
351	from http://www.nature.com/ngeo/journal/v6/n6/full/ngeo1799.html
352	doi: 10.1038/ngeo1799
353	Bony, S., Stevens, B., Coppin, D., Becker, T., Reed, K. A., Voigt, A., & Medeiros,
354	B. (2016, September). Thermodynamic control of anvil cloud amount.
355	Proceedings of the National Academy of Sciences, 113(32), 8927–8932. Re-
356	trieved 2017-10-10, from http://www.pnas.org/content/113/32/8927 doi:
357	10.1073/pnas.1601472113
358	Bony, S., Stevens, B., Frierson, D. M. W., Jakob, C., Kageyama, M., Pincus, R.,
359	Webb, M. J. (2015, March). Clouds, circulation and climate sensi-
360	tivity. Nature Geoscience, 8(4), ngeo2398. Retrieved 2017-11-10, from
361	https://www.nature.com/articles/ngeo2398 doi: 10.1038/ngeo2398
362	Chepfer, H., Bony, S., Winker, D., Cesana, G., Dufresne, J. L., Minnis, P.,
363	Zeng, S. (2010, February). The GCM-Oriented CALIPSO Cloud Prod-
364	uct (CALIPSO-GOCCP). Journal of Geophysical Research: Atmospheres,
365	115(D4), D00H16. Retrieved 2017-11-15, from http://onlinelibrary.wiley
366	.com/doi/10.1029/2009JD012251/abstract doi: $10.1029/2009JD012251$
367	Choi, YS., Kim, W., Yeh, SW., Masunaga, H., Kwon, MJ., Jo, HS., &
368	Huang, L. (2017, June). Revisiting the iris effect of tropical cirrus clouds
369	with TRMM and A-Train satellite data. Journal of Geophysical Re-
370	search: Atmospheres, 122(11), 2016JD025827. Retrieved 2018-02-11, from
371	http://onlinelibrary.wiley.com/doi/10.1002/2016JD025827/abstract
372	doi: 10.1002/2016JD025827
373	Cronin, T. W., & Wing, A. A. (2017, December). Clouds, Circulation, and Climate
374	Sensitivity in a Radiative-Convective Equilibrium Channel Model. Journal of
375	Advances in Modeling Earth Systems, $9(8)$ , 2883–2905. Retrieved 2018-03-
376	09, from http://onlinelibrary.wiley.com/doi/10.1002/2017MS001111/
377	abstract doi: $10.1002/2017MS001111$
378	Del Genio, A. D., & Kovari, W. (2002, September). Climatic Properties of
379	Tropical Precipitating Convection under Varying Environmental Condi-
380	tions. Journal of Climate, 15(18), 2597–2615. Retrieved 2018-03-09, from
381	https://journals.ametsoc.org/doi/abs/10.1175/1520-0442(2002)015%
382	<b>3C2597:CPOTPC%3E2.0.C0%3B2</b> doi: 10.1175/1520-0442(2002)015(2597:

Eitzen, Z. A., Xu, K.-M., & Wong, T. (2009, November). Cloud and Radiative

CPOTPC>2.0.CO:2

383

385	Characteristics of Tropical Deep Convective Systems in Extended Cloud Ob-
386	jects from CERES Observations. Journal of Climate, 22(22), 5983–6000.
387	Retrieved 2019-07-25, from https://journals.ametsoc.org/doi/full/
388	10.1175/2009JCLI3038.1 doi: 10.1175/2009JCLI3038.1
389	Gage, K. S., & Reid, G. C. (1986). The tropical tropopause and the El
390	Niño of 1982–1983. Journal of Geophysical Research: Atmospheres,
391	91(D12), 13315–13317. Retrieved 2020-05-20, from https://agupubs
392	.onlinelibrary.wiley.com/doi/abs/10.1029/JD091iD12p13315 (_eprint:
393	https://agupubs.onlinelibrary.wiley.com/doi/pdf/10.1029/JD091iD12p13315)
394	doi: 10.1029/JD091iD12p13315
395	Harrop, B. E., & Hartmann, D. L. (2012, March). Testing the Role of Radiation
396	in Determining Tropical Cloud-Top Temperature. Journal of Climate, 25(17).
397	5731-5747. Retrieved 2019-06-18. from https://journals.ametsoc.org/doi/
398	full/10.1175/JCLI-D-11-00445.1 doi: 10.1175/JCLI-D-11-00445.1
399	Hartmann, D. L. (2016, August). Tropical anvil clouds and climate sensitiv-
400	ity. Proceedings of the National Academy of Sciences, 113(32), 8897–8899.
401	Retrieved 2019-12-04, from http://www.pnas.org/lookup/doi/10.1073/
402	pnas.1610455113 doi: 10.1073/pnas.1610455113
403	Hartmann, D. L., & Larson, K. (2002, October). An important constraint on trop-
404	ical cloud - climate feedback. <i>Geophysical Research Letters</i> , 29(20), 1951.
405	Retrieved 2017-10-10, from http://onlinelibrary.wilev.com/doi/10.1029/
406	2002GL015835/abstract_doi: 10.1029/2002GL015835
407	Hartmann D. L. & Michelsen M. L. (2002 February) No evidence for iris
408	Bulletin of the American Meteorological Society, 83(2), 249–254.
400	trieved 2018-03-09 from https://journals.ametsoc.org/doj/abs/
410	10.1175/1520-0477(2002)083%3C0249%3ANEFT%3E2.3.C0:2 doi: 10.1175/
411	1520-0477(2002)083(0249:NEFI)2.3.CO:2
412	Hersbach H Bell B Berrisford P Horányi A Muñoz Sabater J Nicolas J
413	others (2019, April). Global reanalysis: goodbye ERA-Interim, hello ERA5.
414	ECMWF Newsletter, No. 159, ECMWF. ECMWF Newsletter(159), 17–24.
415	Jensen, E. J., Read, W. G., Mergenthaler, J., Sandor, B. J., Pfister, L.,
416	& Tabazadeh, A. (1999). High humidities and subvisible cir-
417	rus near the tropical tropopause. Geophysical Research Letters,
418	26(15), 2347–2350. Retrieved 2020-05-13, from https://agupubs
419	.onlinelibrary.wiley.com/doi/abs/10.1029/1999GL900266 (_eprint:
420	https://agupubs.onlinelibrary.wiley.com/doi/pdf/10.1029/1999GL900266) doi:
421	10.1029/1999GL900266
422	Jensen, E. J., Toon, O. B., Selkirk, H. B., Spinhirne, J. D., & Schoeberl, M. R.
423	(1996). On the formation and persistence of subvisible cirrus clouds
424	near the tropical tropopause. Journal of Geophysical Research: Atmo-
425	spheres, 101 (D16), 21361–21375. Retrieved 2020-05-13, from https://
426	agupubs.onlinelibrary.wiley.com/doi/abs/10.1029/95JD03575 (_eprint:
427	https://agupubs.onlinelibrary.wiley.com/doi/pdf/10.1029/95JD03575) doi:
428	10.1029/95JD03575
429	Khairoutdinov, M., & Emanuel, K. (2013). Rotating radiative-convective equilib-
430	rium simulated by a cloud-resolving model. Journal of Advances in Modeling
431	Earth Systems, $5(4)$ , $816-825$ . Retrieved 2019-11-25, from https://agupubs
432	.onlinelibrary.wiley.com/doi/abs/10.1002/2013MS000253 doi: 10.1002/
433	
434	Kubar, I. L., Hartmann, D. L., & Wood, R. (2007, November). Radiative and
435	Convective Driving of Tropical High Clouds. Journal of Climate, 20(22), 5510–
436	5520. Activeved 2019-07-25, from https://journals.ametsoc.org/dol/full/
437	$10 1175 / 0007 101 11609 1 d_{0}; 10 1175 / 0007 101 11609 1$
	10.1175/2007 JCLI1628.1 doi: 10.1175/2007 JCLI1628.1
438	10.1175/2007 JCLI1628.1 doi: 10.1175/2007 JCLI1628.1 Li, Y., Yang, P., North, G. R., & Dessler, A. (2012, February). Test of the Fixed

440	2317-2328. Retrieved 2019-07-25, from https://journals.ametsoc.org/doi/
441	full/10.1175/JAS-D-11-0158.1 doi: 10.1175/JAS-D-11-0158.1
442	Lindzen, R. S., Chou, MD., & Hou, A. Y. (2001, March). Does the Earth Have an
443	Adaptive Infrared Iris? Bulletin of the American Meteorological Society, 82(3),
444	417-432. Retrieved 2018-02-13, from https://journals.ametsoc.org/doi/
445	abs/10.1175/1520-0477%282001%29082%3C0417%3ADTEHAA%3E2.3.C0%3B2
446	doi: 10.1175/1520-0477(2001)082(0417:DTEHAA)2.3.CO;2
447	Liu, R., Liou, KN., Su, H., Gu, Y., Zhao, B., Jiang, J. H., & Liu, S. C. (2017).
448	High cloud variations with surface temperature from 2002 to 2015: Contribu-
449	tions to atmospheric radiative cooling rate and precipitation changes. Journal
450	of Geophysical Research: Atmospheres, 122(10), 5457–5471. Retrieved 2019-
451	00-25, from https://agupubs.onlinelibrary.wiley.com/doi/abs/10.1002/
452	2016JD026303 doi: $10.1002/2016JD026303$
453	Lu, J., Chen, G., & Frierson, D. M. W. (2008, November). Response of the
454	Zonal Mean Atmospheric Circulation to El Nino versus Global Warm-
455	ing. Journal of Climate, 21(22), 5835–5851. Retrieved 2020-05-20, from
456	https://journals.ametsoc.org/doi/full/10.1175/2008JCL12200.1 (Pub-
457	Mauritary T. & Charge D. (2015 Mar). Missing init affect on a marilla course of
458	Mauritsen, I., & Stevens, B. (2015, May). Missing iris effect as a possible cause of
459	seign as $\Re(5)$ 246 251 Betrieved 2010 02 07 from https://www.neture.com/
460	science, $\delta(5)$ , $540-551$ . Refleved 2019-02-07, from fittps://www.flature.com/
461	Morice C D Kennedy I I Bayner N A & Jones D D (2012 April) Quanti
462	fring uncertainties in global and regional temporature change using an ensem
403	ble of observational estimates: The HadCRUT4 data set: THE HADCRUT4
404	DATASET Journal of Geonbusical Research: Atmospheres 117(D8) n/a-n/a
405	Retrieved 2019-10-28 from http://doi.wilev.com/10.1029/2011.ID017187
467	doi: 10.1029/2011.ID017187
468	Stephens G Winker D Pelon J Trepte C Vane D Yuhas C Lebsock
469	M. (2017, August). CloudSat and CALIPSO within the A-Train: Ten years of
470	actively observing the Earth system. Bulletin of the American Meteorological
471	Society. Retrieved 2018-03-07, from https://journals.ametsoc.org/doi/
472	abs/10.1175/BAMS-D-16-0324.1 doi: 10.1175/BAMS-D-16-0324.1
473	Su, H., Jiang, J. H., Neelin, J. D., Shen, T. J., Zhai, C., Yue, Q., Yung, Y. L.
474	(2017, June). Tightening of tropical ascent and high clouds key to precipita-
475	tion change in a warmer climate. Nature Communications, 8, 15771. Retrieved
476	2019-01-24, from https://www.nature.com/articles/ncomms15771 doi:
477	10.1038/ncomms15771
478	Thompson, D. W. J., Bony, S., & Li, Y. (2017, August). Thermodynamic constraint
479	on the depth of the global tropospheric circulation. Proceedings of the National
480	Academy of Sciences, 114(31), 8181-8186. Retrieved 2018-02-11, from http://
481	www.pnas.org/content/114/31/8181 doi: 10.1073/pnas.1620493114
482	Tompkins, A. M., & Craig, G. C. (1999, February). Sensitivity of Tropical
483	Convection to Sea Surface Temperature in the Absence of Large-Scale
484	Flow. Journal of Climate, $12(2)$ , $462-476$ . Retrieved 2018-02-23, from
485	https://journals.ametsoc.org/doi/full/10.1175/1520-0442(1999)
486	012%3C0462:S0TCTS%3E2.0.C0%3B2 doi: 10.1175/1520-0442(1999)012(0462:
487	SOTCTS>2.0.CO;2
488	Wang, PH., McCormick, M. P., Poole, L. R., Chu, W. P., Yue, G. K., Kent, G. S.,
489	& Skeens, K. M. (1994, June). Tropical high cloud characteristics derived from
490	SAGE II extinction measurements. Atmospheric Research, 34(1), 53–83. Re-
491	trieved 2020-05-13, from http://www.sciencedirect.com/science/article/
492	$p_{11}/0109809594900817  doi: 10.1010/0109-8095(94)90081-7$
493	[Data gat] NASA Longley Atmentic Gringer D. J. G. J. DAAG 1: 10
494	[Data set]. NASA Langley Atmospheric Science Data Center DAAC doi: 10

495	.5067/CALIOP/CALIPSO/L3_CLOUD_OCCURRENCE-STANDARD-V1-00
496	Winker, Chepfer, H., Noel, V., & Cai, X. (2017, November). Observational
497	Constraints on Cloud Feedbacks: The Role of Active Satellite Sensors.
498	Surveys in Geophysics, 38(6), 1483–1508. Retrieved 2018-03-09, from
499	https://link.springer.com/article/10.1007/s10712-017-9452-0 doi:
500	10.1007/s10712-017-9452-0
501	Winker, Hunt, W. H., & McGill, M. J. (2007, October). Initial performance as-
502	sessment of CALIOP. Geophysical Research Letters, 34(19), L19803. Re-
503	trieved 2018-03-07, from http://onlinelibrary.wiley.com/doi/10.1029/
504	2007GL030135/abstract doi: $10.1029/2007$ GL030135
505	Xu, KM., Wong, T., Wielicki, B. A., Parker, L., & Eitzen, Z. A. (2005, July).
506	Statistical Analyses of Satellite Cloud Object Data from CERES. Part
507	I: Methodology and Preliminary Results of the 1998 El Niño/2000 La
508	Niña. Journal of Climate, 18(13), 2497–2514. Retrieved 2019-07-25, from
509	https://journals.ametsoc.org/doi/full/10.1175/JCLI3418.1 doi:
510	10.1175/JCLI3418.1
511	Xu, KM., Wong, T., Wielicki, B. A., Parker, L., Lin, B., Eitzen, Z. A., & Bran-
512	son, M. (2007, March). Statistical Analyses of Satellite Cloud Object Data
513	from CERES. Part II: Tropical Convective Cloud Objects during 1998 El
514	Niño and Evidence for Supporting the Fixed Anvil Temperature Hypoth-
515	esis. Journal of Climate, $20(5)$ , $819-842$ . Retrieved 2019-07-25, from
516	https://journals.ametsoc.org/doi/full/10.1175/JCLI4069.1 doi:
517	10.1175/JCLI4069.1
518	Yuan, J., Houze, R. A., & Heymsfield, A. J. (2011, March). Vertical Structures of
519	Anvil Clouds of Tropical Mesoscale Convective Systems Observed by CloudSat.
520	Journal of the Atmospheric Sciences, 68(8), 1653–1674. Retrieved 2019-12-04,
521	from https://journals.ametsoc.org/doi/full/10.1175/2011JAS3687.1
522	doi: 10.1175/2011JAS3687.1
523	Zelinka, M. D., & Hartmann, D. L. (2010, August). Why is longwave cloud feedback
524	positive? Journal of Geophysical Research: Atmospheres, 115(D16), D16117.
525	Retrieved 2017-11-10, from http://onlinelibrary.wiley.com/doi/10.1029/
526	2010JD013817/abstract doi: 10.1029/2010JD013817
527	Lemnka, M. D., & Hartmann, D. L. (2011, December). The observed sensitivity of
528	ingle clouds to mean surface temperature anomalies in the tropics. Journal of $C_{\text{combassion}}$ Research: Atmospheres, $116(D22)$ , D22102, Detriver 1,2017,11
529	Geophysical Research: Atmospheres, 116(D23), D23103. Retrieved 2017-11-
530	U5, from http://onlinelibrary.wiley.com/doi/10.1029/2011JD016459/

abstract doi: 10.1029/2011JD016459