Observing convective activities in the complex <u>convective</u> organizations and their contributions to the precipitation and anvil <u>cloud</u> amounts

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Abstract. The <u>convective</u> processes of convection precipitating precipitation and producing the production of anvil clouds determine the Earth's water and radiative budgets. However, convection could have very complicated <u>convective</u> organizations and behaviors in the tropics. A bunch of<u>Many</u> convective activities of <u>in</u> various life stages would beare connected together in the complex <u>convective</u> organizations, and it is difficult to distinguish their behaviors, e.g., precipitating, producing the anvil, <u>merging and splitting</u>. In this work, from <u>on</u> the <u>basis of</u> hourly satellite images of the infrared brightness temperature (BT) satellite images, with a novel variable-BT tracking algorithm, complex convective organizations are partitioned into organization segments of single cold cores as tracking targets, the organization segments of a single but variable BT cold core <u>Tare identified and tracked</u>. By the segment tracking, the detailed evolution of the organization and anvil <u>clouds are explicitly</u> associated with unique cold cores. Compared with previous tracking algorithms that focused only on variations in areas, the <u>novel variable-BT</u> tracking algorithm is capable of documenting the evolution of both the area and BT structures. contributions For validation, the tracked motions are compared against the radiosonde cloud-top winds, with mean speed differences of -1.6 m/s and mean angle differences of 0.5°.

- 20 With the novel variable-BT tracking algorithm, the behaviors of oceanic convection over the tropical western Pacific Ocean are investigated. of each organization segment are distinguished from the connected convection complex. The results show that the-duration, precipitation and anvil amount of the lifecycle accumulation all of the tracked organization segments have a-simple log-linear relationships with the cold-core-peak BT. The organization segments of the peak BT values less than 220 K are long-lived, with average durations of 4-16 hours, whereas the core colder than 220K are the most robust with the
 - 25 duration of 4-16 hours, while the organization segments of the shallow-warmer-peak BT values -structures-disappear rapidly within a few hours but with a high occurrence frequency. The decay process after the cold core peaks contributes to more precipitation and anvil clouds than does the development process. With the core peaking at a colder BT, the differences in the accumulated duration, precipitation and anvil production between the development and decay stages increase exponentially. Additionally, the disappear rapidly in a few hours but are the most frequent. The occurrence frequency of the mergers and splits
 - 30 also increases exponentially with the decrease of has a loglinear relationship with the cold-core-peak BT. For the lifecycles of the same cold-core-peak BT, the lifetime-accumulated precipitation and anvil amount are strongly enhanced in complicated lifecycles with the occurrence of mergers and splits compared with those with no mergers or splits. For the total tropical convective cloud water budget, long-lived complicated lifecycles make the largest contribution to precipitation, whereas long-lived complicated and short-lived simple lifecycles make comparable contributions to the anvil cloud amount and are both
 - 35 <u>important.By the mergers and splits, more high cloud systems are born from convection and the lifetime accumulated precipitation and anvil amount are strongly enhanced as compared to those of no mergers and splits. Overall, 85.4% of tropical precipitation are contributed by the long lived organization segments, in which 67.7% are accompanied with mergers or splits. The tropical non precipitating anvil amount are mostly contributed by both long lived organization segments or splits.</u>

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1. Introduction

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Precipitation and anvil clouds are two key components of the convective cloud water budget but are usually accompanied with by very complicated microphysical and dynamic processes. In climate models, their representations is are determined by tunable parameters of with large uncertainty uncertainties, e.g., the detrainment and precipitation efficiency (Rennó et al., 1994; Zhao, 2014; Clement and Soden, 2005; Zhao et al., 2016; Suzuki et al., 2013). In cloud-resolving models, the -parameterization scheme of convection is still subject to many uncertainties in ice microphysics and sub-grid turbulence (Matsui et al., 2009; Blossey et al., 2007; Powell et al., 2012; Bretherton, 2015; Atlas et al., 2024), although cloud dynamics and microphysics can be resolved at fine scales. The challenge is partially due tobecause that the detailed convection processes of convection precipitatingprecipitation and producing the production of anvil clouds -- have not been sufficiently explored from the observations to advance understanding and model parameterization.

The spatial organization of convection varies from a simple isolated cell to the a complex complicated -structure that consists of a bunch of many convective activities of in various life stages. and Tits he variation in convection organizations is closely related to the changes of thein precipitation and the anvil cloud amount (Yuan and Houze, 2010; Yuan et al., 2011; Tobin et al., 2012; Wing and Emanuel, 2014; Mauritsen and Stevens, 2015; Ruppert and Hohenegger, 2018; Bony et al., 2020; 55 Bao and Sherwood, 2019; Houze, 2004). Cloud-resolving models and observations both suggest that the convective organizations are the important bridges of for the interactions between convection and the environment (Tobin et al., 2012; Blossey et al., 2005; Coppin and Bony, 2015; Wing and Emanuel, 2014; Wing et al., 2017; Holloway et al., 2017; Muller and Bony, 2015; Sokol and Hartmann, 2022). By the Through radiative feedback and circulation, the convective organizations is are associated with the nonconvecting environment. The dDrier free troposphere and enhanced radiative cooling of the 60 nonconvecting regions can would reinforce the subsidence to expand the dry region and thereby force the convection of in the moist region to aggregate (Blossey et al., 2005; Coppin and Bony, 2015). Over the-warm oceans, stronger mass convergence and surface turbulent fluxes would promote the aggregation by developing the deep convection and inhibiting scattered convective activities (Coppin and Bony, 2015; Holloway et al., 2017; Wing et al., 2017). Although organizational The organization-variations -can influence the precipitation efficiency (Bao and Sherwood, 2019) (Lindzen et al., 2001; Mauritsen 65 and Stevens, 2015; Choi et al., 2017), but under the condition when the total atmospheric water amount is not known and is difficult to measure, the increased precipitation efficiency does not guarantee the a decrease of in the anvil cloud amount. Thus, the links between among convective organizations, precipitation and anvils anvil clouds still need further observational evidences as constraints for understanding their climate feedback processes.

However, oObserving the organizations and behaviors of convection is still challenging. Although active radar and lidar 70 sensors from on polar-orbit satellites or and ground-based observatories can penetrate convection convective clouds, their spatiotemporal sampling is too sparse for tracking. From the images of the brightness temperature at 10.8 µm (BT₁₁) of geostationary satellites (GEOs), pixels of thin cirrus clouds cannot be accurately distinguished from cloudless pixels, but the major structure of the organized convection, consisting of the deep convective clouds and the associated anvil clouds, can be observed continuously in time and used for tracking pixels of thin cirrus cannot be distinguished from cloudless pixels but the 75 organized convective structures with two distinct modes of deep convection and anvils can be well captured and tracked (Richards and Arkin, 1981; Hendon and Woodberry, 1993; Fu et al., 1990). For the identification of convection of convection from GEO images, two methods have been proposed in previous studies. One method is to identify the contiguous area under a fixed threshold of BT11 threshold (Goyens et al., 2011; Schröder et al., 2009; Huang et al., 2018; Williams and Houze, 1987; Chen and Houze, 1997; Kolios and Feidas, 2009; Laing et al., 2008; Feidas and Cartalis, 2007; Fu et al., 2023; Yang et al., 2020; Tsakraklides and Evans, 2003). By On the basis of the fixed threshold, the identified targets are usually of

complex<u>organizations</u> of connected convection is usually identified to track and it is hard tonot capable of being used to distinguish the detailed convective activities and their contributions to the precipitation and anvils<u>inside complex</u> organizations in the complex. Besides<u>In addition</u>, the variable-BT₁₁ identification has been proposed in recent years, in which by applying the <u>a set of</u> adaptive BT₁₁ thresholds <u>are used</u> to divide the clustered convection complex into independent convective systems for tracking (Yuan and Houze, 2010; Fiolleau and Roca, 2013; Feng et al., 2023; De Laat et al., 2017; Bouniol et al., 2016; Heikenfeld et al., 2019; Zinner et al., 2008; Zinner et al., 2013). This approach brings the possibility<u>makes</u> it possible to track the detailed variations of <u>in complex</u> the convection organizations but has not been fully achieved, particularly on the aspect of with respect to tracking the evolution of the <u>3-D-BT₁₁</u> structures.

- Convective systems could can merge and split and their BT₁₁BT structures could can change rapidly. These complicated
 behaviors make them difficult to track. To determine the life associations between convective systems, oTne of the most applicable-widely used tracking methods is based on the area-overlap in areas between two targets at different timesping rates (Williams and Houze, 1987). This method permits the mergers and splits but has flaws in tracking fast-moving clouds (Huang et al., 2018). On the other hand, Besides, to track the cloud movements, the most widely used method is based on can be well derived via the cross correlation to match the cloud BT₁₁ patterns at different times (Leese et al., 1971; Nieman et al., 1997;
 Velden et al., 1998; Salonen and Bormann, 2016; Hersbach et al., 2020). Thus, tThoese two methods are complementary and usually-can be combined together to first derive cloud displacements and then determine the temporal associations between targets according to the dynamic overlaps (i.e., the overlap in areass after the movements) (Feng et al., 2023; Zinner et al., 2013).
- In this work, the complex <u>convective convection</u> organizations (CCOs) are segmented <u>into simple structural components</u> of single cold cores and tracked separately according to variable-BT₁₁ identification and dynamic overlap. Compared with fixed-threshold tracking, the variable-BT₁₁ tracking algorithm has the advantages of documenting more detailed convective evolution in CCOs. Although several variable-BT₁₁ tracking algorithms have been proposed, the tracked lifecycle is still described mostly by the variation in areas and lacks of the BT₁₁ structural information. By the novel variable-BT₁₁ tracking algorithm developed in this work, the tracked lifecycle is described by the cold-core BT₁₁ variation in the CCO structural components. The precipitation and non-precipitating anvil clouds are explicitly associated with unique cold cores. and tracked based on the variable BT₁₁ identification and dynamic overlaps.

In comparison to the previous tracking algorithms, this work more focuses on the evolution on the BT₁₁-dimension. By tracking the organization segments, the aims of this work are twofold: (1) what is the dependence of the duration, precipitation and anvil amount of organization segments on the BT₁₁-structures? (2) what are the contributions of organization segments to the tropical precipitation and anvil amount? This paper is laid out as follows: Sect. 2 describes the data and methods used in our analyses; Sect. 3 introduces the <u>novel</u> variable-BT₁₁ segment tracking algorithm and its comparison to <u>fthe fixed-threshold</u> tracking. Sect. 4 explores the relationships of <u>the convective</u> duration, precipitation and anvil amount. <u>Sect.</u> 5 presents conclusions.

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2. Data and methods

2.1 Images from GEOs

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<u>The</u> BT₁₁ in the tropics between 20°S-20°N and 90°E<u>W</u>-190170°EW was scanned by radiometer imagerss on the geostationary Multi-functional Transport Satellite 1 Replacement and 2 Replacement (MTSAT-1R and -2R), with a scanning start time of s at the half an hour and a-view zenith angles of less than 60°. The BT₁₁ images from from 2006 with -1-hour and 8-km resolutions are included in were obtained from the Satellite ClOud and Radiative Property retrieval System (SatCORPS) of the Clouds and the Earth's Radiant Energy System (CERES) project. In the CERES SatCORPS, the BT₁₁ was calibrated

against the Moderate Resolution Imaging Spectroradiometer (MODIS) from the Aqua (Doelling et al., 2013; Doelling et al., 2016). To facilitate the data processing, the BT₁₁ images of 8-km pixels were were further gridded to 0.05° by via linear interpolation (Amidror, 2002).

2.2 Global precipitation measurement (GPM)

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At fine scales (0.1° and half-hour resolution), the GPM data combines all available sensors for precipitation estimates, including microwave imagers from multiple low-Earth orbit satellites, the infrared (IR) channel of GEO radiometers and landsurface rain gauges (Huffman et al., 2007; Huffman et al., 1997). The microwave brightness temperature is sensitive to 130 atmospheric hydrometers for of precipitation but has sparse spatiotemporal sampling due to the its sun-synchronous orbit. For grid boxes without microwave observations, the GEO-IR BT₁₁ is was used to estimate precipitation according to the spatially varying calibration coefficient of the microwave precipitation rates (Huffman et al., 1997). To improve the accuracy, rain gauges are-were further used to rescale the-satellite estimates of precipitation rates (Huffman et al., 1997). It has been demonstrated that this satellite-based precipitation product performs well for strong precipitation events with the a mean bias .35 smaller than 1_-mm/day but misses 20-80% of the light precipitation (< 10_-mm/day) (Tian et al., 2009). In the tropics, light precipitation (< 1_-mm/hour) accounts for approximately 55-70% of the precipitation area but contributes to only 9-18% of the total precipitation (Yuan and Houze, 2010). Only precipitation rates greater than 1 mm/hour are considered the regions with precipitation in this work since light precipitation has high uncertainty in the GPM data and relatively low contributions to total precipitation.

Given the large uncertainty in the detection of satellite based light precipitation but the relatively low importance of total precipitation, only precipitation rates greater than 1 mm/hour are considered as precipitating regions in this work.

2.3 Cloud-top winds from m-ground-based radar and radiosonde observations

The cCloud detection from radar and wind observations of the radar and from radiosondes are were combined to derive the cloud-top winds for at three tropical ground-based observatories of the Atmospheric Radiation Measurement (ARM) 45 program:-at Darwin (130.9°E, 12.4°S), Manus Island (147.4°E, 2.1°S) and Nauru Island (166.9°E, 0.5°S). The vertical distribution of hydrometers up to 200 km above the ground is-was detected byvia -35 GHz millimeter-wave cloud radar (MMCR), with temporal and spatial sampling resolution of 10-s and 45-m, respectively. The best estimate reflectivity of the MMCR in the range of -50 to 20 -dBZ is-was provided in the ARM program Active Remote Sensing of Clouds (ARSCL) value-added product for at the three sites. A reflectivity higher than -40_-dBzZ corresponds to was identified as a clouds (Zhao .50 et al., 2017). To match the GEO observations, the instantaneous cloud profiles within 5 minutes around the time of the GEO radiometer imager scanning at time these three sites of the GEO imagers were were collected at those three sites to compute the cloud fraction, respectively. The 10-min cloud-fraction profile is-was computed as the ratio of the number of cloud occurrences to the total number of observations at each height. Continuous levels of the cloud fraction greater than zero were identified as cloud layers. Cloud layers are identified as layers in which the fraction exceeds zero. The thickest high cloud .55 layer with the a top higher-greater than 5 -km and a maximum cloud fraction of at least 50% is was selected as the major high cloud layer passing over the sites to collocate with the high clouds observed by GEOs. The cloud top refers to the uppermost height of the major cloud layer with the fraction greater than zero.

Winds are-were_detected by the ARM balloon-borne radiosondes with high vertical and temporal resolutions of 10_-m and 2.5_-s, respectively. The accuracy of the radiosonde wind speed is was approximately 0.5_-m/s. The radiosondes are were 160 launched two times a day at Manus and Nauru (approximately 11:30 and 23:30 UTC) and four times a day at Darwin (approximately 4:30, 11:15, 16:30 and 23:15 UTC). In comparisonCompared with the instantaneous cloud detection via of the MMCR, balloon-borne radiosondes might take hours to-and drift dozens of kilometers away from the launch location to approach the upper tropospherecloud top height. To collocate the cloud and wind observations, the time difference between between the time of cloud top detection and the time of the balloon-borne radiosondes reaching the cloud top and the MMCR observations of the cloud top must be within one hour for quality control. To match with the tracked motions of the organization

segments, the radiosonde cloud top winds within 150 km of the segment centroids are considered as the observational reference and the selection of this distance is consistent with previous studies examining the performance of cloud drift winds (Nieman et al., 1997; Santek et al., 2019; Daniels et al., 2020).

2.4 Comparison of the cloud-top winds and the tracked cloud motions

The difference between the <u>observational_observed</u> cloud-top winds and the tracked <u>cloud</u> motions is assessed-<u>as</u> <u>followsby the mean speed and angle bias, the mean vector difference (MVD), the standard deviation (SD) of the MVD and the</u> root mean square error (RMSE), <u>which is</u> consistent with <u>those in</u> Nieman et al. (1997):

Speed BIASbias =
$$\frac{1}{N} \sum_{i=1}^{N} \left(\sqrt{U_i^2 + V_{i\neq}^2} - \sqrt{U_{ri}^2 + V_r^2} \right),$$
 (1)

Angle bias
$$\frac{BLAS}{N} = \frac{1}{N} \sum_{i=1}^{N} (\arctan\left(\frac{V_i}{U_i}\right) - \arctan\left(\frac{V_r}{U_r}\right)),$$
 (2)

$$MVD = \frac{1}{N} \sum_{i=1}^{N} \sqrt{(U_i - U_r)^2 + (V_i - V_r)^2},$$
(3)

$$SD = \sqrt{\frac{1}{N} \sum_{i=1}^{N} (\sqrt{(U_i - U_r)^2 + (V_i - V_r)^2} - MVD)^2},$$
(4)

$$RMSE = \sqrt{MVD^2 + SD^2}.$$
(5)

Here, the mean speed and angle bias, the mean vector difference (MVD), the standard deviation (SD) of the MVD and the root-mean-square error (RMSE) of the tracked cloud motions compared with the observational cloud-top winds were computed.
 UU and V are the x- and y-component winds, respectively. The subscripts i and r indicate an individual sample of the tracked cloud motion and the corresponding reference cloud-top winds from theof radiosondes observations, respectively, and N is the total number of samples.

2.5 Pattern-matching displacement

The displacement of the organization segment is derived by searching for the maximum similarity of the pattern in the latter image based on cross correlation (Leese et al., 1971; Velden et al., 1998). The target scene is selected as the irregular segment BT₁₁-pattern with shapes and is matched in the later image to find the displacement by minimizing the sum of squared differences (SSD) of the normalized BT₁₁:

$$SSD = \sum_{x,y} [BT_{\pm\pm}^{\prime}(x,y) - \widetilde{BT}_{\pm\pm}^{\prime}(x,y)]^{2}.$$
(6)

190 BT⁺₁₁(x, y) and BT⁺₁₁(x, y) are the normalized BT₁₁ values at pixel (x, y) of the target scene and the cross scene in the search, respectively. By normalizing BT₁₁, the minimum SSD corresponds to the maximum pattern correlation for the BT₁₁ structures. The search region is confined by the displacements smaller than 50 km per hour, which is the maximum motion predicted by models (Merrill et al., 1991). The final match is further examined by the coefficient of the pattern correlation. For the areas larger (smaller) than 5000 km², the match is valid with a pattern correlation higher than 0.6 (0.8), in which the threshold values are consistent with those of Daniels et al. (2020). Otherwise, the BT₁₁ structures would change rapidly and rather be considered to be stationary and the displacement is zero.

2.62.5 *t* test and confidence intervals

The 95% confidence interval for the mean value was computed <u>based onvia</u> the t test: $\bar{x} \pm t_c \frac{s}{\sqrt{N}}$, where, \bar{x} is the mean value of all samples t_c is the critical value for t.—<u>: and s</u> is the standard deviation of all the samples. N is the number of independent samples, which and N is determined by the sample length divided by the distance between independent samples (Bretherton et al., 1999).

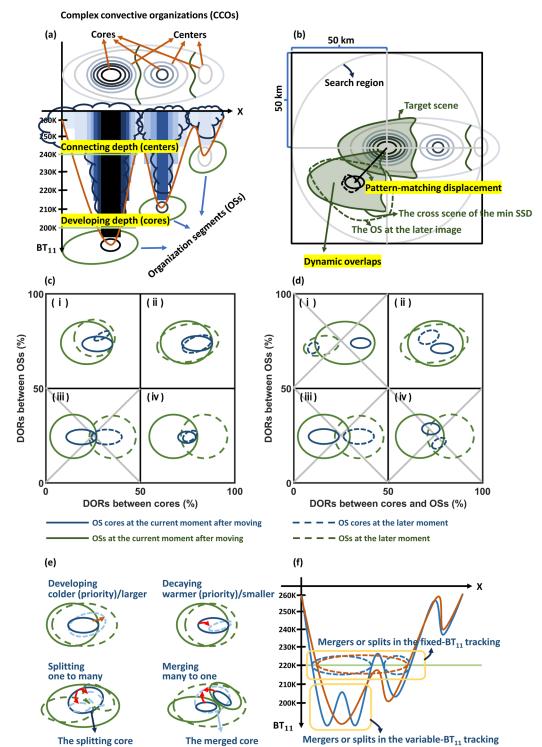
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3. Tracking the convective organization segments



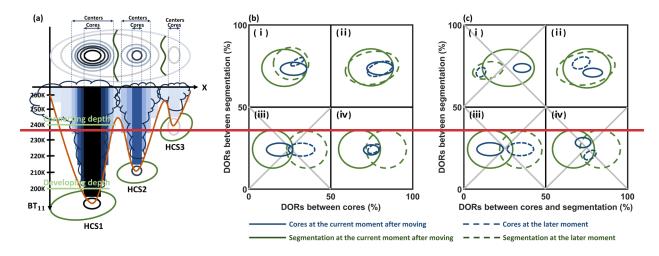


Figure 1. Illustrations of the variable-BT₁₁ segment tracking algorithm. (a) Example illustrations of segmenting the elustered convection complex<u>CCO</u> into single-core three HCSs-OSs as tracking targets. The CCO, whose 3-D-dimensional structures in x, y and BT₁₁ are captured-identified by the adaptive variable-BT₁₁ identificationthresholds. The cold-core BT₁₁ indicates the depth of development. The cold-center BT₁₁ indicates the depth of the connection. (b) Example illustrations of tracking the OS by combining cross correlation and the overlap in areas. The OS is moved byaccording to the displacement predicted by cross correlation and then overlaps with the OSs in the later images. (c-d) The dynamic overlapping situations of two OSs of different moments when their cores have overlaps (c) and no overlaps (d), respectively, respectively. The solid blue and green lines indicate the OS core and segmentation outlines of of_the current moment_at the location predicted by cross correlation, respectively, at the location. The dashed blue and green lines indicate the OS core and segmentation outlines of the denial of the_non_association between OSs. (e) Examples illustrating tracked OS evolution (i.e., development, decay, mergers and splits). The red arrows indicate the OS evolution of the OS with the time. (f) Illustrations of the difference between the variable-BT₁₁ and the-fixed-BT₁₁ tracking for the-mergers and splits. The 220K threshold is taken as an example for the ...The solid red and blue lines are the CCO BT₁₁ structures of at different times captured by the adaptive variable-BT₁₁ thresholds. The dashed red and blues linescontours are the tracking-mergers and splits captured by the fixed threshold of 220 K.

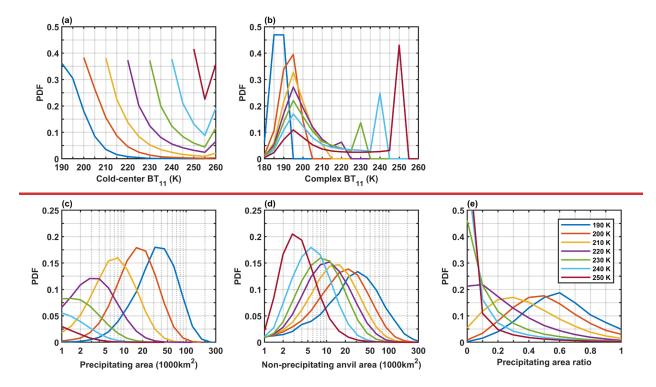


Figure 2. The 3 D structure characteristics of the organization segments. The PDFs of the cold center BT₁₁ (a), complex BT₁₁ (b), precipitating area (c), non-precipitating area (d) and precipitating area ratio (e) of the segmented organization structures with the cold core BT₁₁ from 190 250K.

To distinguish the behaviors of the clustered convective activities in the complex organization \underline{CCOs} , the organization segments of high cloud systems (HCSs)(\underline{OSs}) with of a single but variable- BT_{11} cold cores (Fig. 1a) are identified

230 usedpartitioned as the tracking targets (Fig. 1a) and are further tracked by by combining the cross correlation and the area overlap (Fig. 1b) based on the hourly infrared satellite images. This novel variable-BT₁₁ segment tracking algorithm and its difference from the conventional fixed-threshold tracking algorithm are introduced in this section as follows.

(1) Segmenting sCCOs into the OSs of single cold cores

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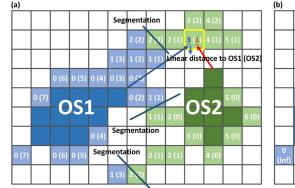
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As illustrated dynamic overlaps (Fig. 1b c) from the hourly infrared satellite images. The details of the novel variable BT₁₁-segment tracking algorithm and its difference to the convectional tracking algorithm are described in this section as follows.

in Fig. 1a, the CCO is the <u>complex organization eonnected eonvection eomplex of multiple connected convections</u> and <u>is</u> identified as the contiguous area of the BT₁₁ that is colder than 260 K. The 260-K threshold can enclose 95% of deep convective clouds and as much of the anvil <u>cloud</u> as possible but with the least contamination from lower-level clouds (Yuan and Houze, 2010; Yuan et al., 2011; Chen and Houze, 1997). The segmented single-core structural component of CCOs is identified as the OS to be used as <u>the</u> tracking targets. A set of adaptively variable-BT₁₁ thresholds from 180-260 K per 5-K interval and the a minimum area threshold of 1000 km² are used to capture the "growth rings" in <u>the clustered convective complex</u> organizationsCCOs. In Fig. 1a, these rings reflect the <u>CCO</u> structure in three dimensions <u>of</u> <u>namely</u> x, y and BT₁₁ and are the fundamental indicators of internal dynamics (Houze, 2004). The innermost ring of the <u>local</u> coldest <u>local</u> BT₁₁ is <u>defined</u> as the cold core <u>of</u>, which is the most active vertically developing region in the <u>HCSsOS</u>. The BT₁₁ of the cold core represents the depth at which the OS developed. For a CCO consisting of multiple cold cores, the BT₁₁ of the coldest core in the CCO is <u>defined as the CCO BT₁₁ for representing the depth of CCO development</u>. The isolated ring of the warmest BT₁₁ and enclosing only one core is_the cold center <u>and</u>. The HCS <u>OS</u> would be connected (disconnected) to the surrounding <u>HCSs-OSs</u> outside (within<u>inside</u>) the center_{as} <u>and Thus thus</u> the cold-center BT₁₁ <u>can indicate can be used to indicate</u> the connecting condition of between the <u>HCS-OSs</u> in the complex organization the CCO.



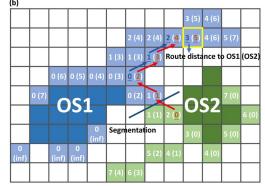


Figure 2. Illustrations of segmentation according to the nearest linear distance (a) and the nearest route distance (b). The dark blue and green pixels represent the OS1 and OS2 centers, respectively. The colored pixels outside the centers are the pixels to be assigned in the contour of the cold-center BT₁₁ plus 1 K. The light blue and green pixels are assigned to OS1 and OS2, respectively. The numbers inside those pixels indicate the number of necessary pixels to connect with OS1 and OS2, respectively. The arrows in (a) and (b) represent the nearest distances of OS1 and OS2 to reach the yellow-edge pixel, as examples to illustrate the computations of the linear distance and the route distance, respectively.

For segmentation, the pPixels lying outside the the centers are assigned to the connected neighborhood HCSs-OSs by
the 1-K interval to achieve segmentation. To be specific, all BT₁₁ contours of the 1-K interval between the cold-center BT₁₁ and 260 K need to be found first. The assignment of the pixels outside the centers is conducted in the order from cold to warm BT₁₁ contours of the 1-K interval. The initial OS is just the center and it is updated after every 1-K-interval assignment. An example illustration of the 1-K-interval assignment is shown in Fig. 2. On the basis of the 8-point-connected neighborhood in which the 8 surrounding points are recognized as the connected neighborhood to the center point, the distance between two pixels is computed as the number of necessary pixels connecting them. According to the nearest linear distance, as shown in Fig. 2a, some of the pixels assigned to OS2 (those light green pixels in Fig. 2a) are disconnected from OS2 but connected to OS1. After the assignment, OS2 is composed of two disconnected parts. For an organized convective system, the assigned

pixels outside the center can also be understood as outflowing anvil clouds from the center. It would be strange that the

- outflowing anvil clouds from OS2 are not connected with its original OS2 but connected with OS1. To avoid these conditions, 270 the distance of the nearest route is used to determine the pixel assignment. Here, the route of OS1 and OS2 to reach a pixel (the blue and red arrows in Fig. 2b) is confined to within the 1-K-interval contour. Pixels of the same distance to OS1 and OS2 are randomly assigned. In Fig. 2b, the assignment of the pixels on the basis of the distance of the nearest route is more reasonable than that in Fig. 2a on the basis of the nearest linear distance. Thus, in every 1-K-interval assignment, the distance of the nearest route is used to accomplish the segmentation and the OSs are updated with these newly assigned pixels iteratively until all the pixels within the CCO are assigned.
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-<u>T</u>The <u>final segments OS have is the 3-D-dimensional (x, y and BT₁₁)</u> structures of <u>a a single but variable BT₁₁-cold core.</u> The OS can be further separated into precipitating and non-precipitating (precipitation less than 1 mm/hour) regions on the basis of the GPM. The non-precipitating area is identified as the anvil cloud. By segmentation, those precipitation and anvil pixels are explicitly associated with unique cold cores. The key definitions of variable-BT11 tracking are depicted in Fig. 1a and summarized in Table 1 for easy checking, and its contribution to the precipitation and non precipitating anvil area can be distinguished from the clustered convective complex organization by the segmentation.

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Name	Definition
Complex convective organizations (CCOs)	The contiguous area of the BT ₁₁ colder than 260 K.
Organization segments (OSs)	The segmented single-core structural component of CCOs.
Cold-core BT ₁₁ (OS developing depth)	The local coldest BT ₁₁ contour in OSs.
Cold-center BT ₁₁ (OS connecting depth)	The local warmest isolated BT ₁₁ contour of only enclosing one core in OSs
CCO BT ₁₁ (CCO developing depth)	The coldest cold-core BT ₁₁ of multiple cores in the CCO.
Anvil cloud	The non-precipitating (precipitation less than 1 mm/hour) region of each
	<u>OS.</u>
Dynamic overlapping rates (DORs)	The OS is moved to the location predicted by cross correlation and then
	overlaps with the OSs in the later image.
Merger and split BT ₁₁	The BT ₁₁ of the merged cold core and the BT ₁₁ of the splitting cold core.
Cold-core-peak BT ₁₁	The coldest cold-core BT ₁₁ in lifecycles, representing the convective
	peaking strength.
Development and decay stages	The stage before and after the time of the cold core peaking at the coldest
	BT ₁₁ (if there are multiple cores of the same BT ₁₁ , the one of the largest co
	areas is selected).
Lifecycle-accumulated duration,	The accumulated time, precipitation and anvil cloud amount in the lifecyc
precipitation and anvil cloud amount	

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(2) Tracking the displacement of OSs on the basis of cross correlation

In Fig. 2a, according to the probability distribution functions (PDFs) of the cold center BT₁₁, it is the most frequent about 35-45% that the cold center BT₁₁ is the same as the cold core BT₁₁ and the isolated deep convective body in the 260 K shield is rare. The isolated structure is less frequent but seems to be another mode for shallow warm systems. Fig. 2b shows the PDFs of the complex BT₁₁ that refers to the coldest cold core BT₁₁ in the connected complex. It can be seen that the HCSs of the cold cores of 200 230K are the most frequently clustered in the 195 K complex, and the HCSs of the cores of 220 250K have two modes for the elustered 195 K complex and the isolated shallow warm structures, respectively. This implies the elustered complex organizations are the major mode for the deep convection and thus the segment tracking is necessary to distinguish their behaviors in the complex organizations. By segmentation, Figs. 2e-e show the PDFs of the HCS precipitating and anvil area basically conform to the log-normal distribution and are closely related to the cold-core BT₁₁. The HCSs of colder cores normally contribute to larger precipitating and anvil areas (Fig. 2e-d) but would be more dominated by the precipitation with higher ratios of the precipitating area to the total HCS area (Fig. 2e). The OS displacement is derived by searching for the maximum similarity of its BT₁₁ pattern in the later image on the basis of the cross correlation (Leese et al., 1971; Velden et al., 1998). As shown in Fig. 1b, the target scene is the OS BT₁₁ pattern. The search region is centered at the core centroid of the target and confined to a radius of 50 km, which corresponds to a maximum OS motion of 50 km/hour (Merrill et al., 1991). The cross scene has the same shape as the OS target and refers to all possible scenes to match the OS target within the search region. The BT₁₁ pattern of the target scene is normalized, and so is the BT₁₁ pattern of each cross scene. The pattern-matching displacement is determined by the minimum of the sum of squared differences (SSD) of the normalized BT₁₁ between the OS target scene and the cross scene.

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$$SSD = \sum_{x,y} [BT'_{11}(x,y) - \widetilde{BT}'_{11}(x,y)]^2,$$
(6)

where $BT'_{11}(x, y)$ and $\widetilde{BT}'_{11}(x, y)$ are the normalized BT_{11} values at pixel (x, y) of the target scene and the cross scene in the search, respectively. Here, the minimum SSD corresponds to the maximum pattern correlation. The final match is examined by the pattern correlation coefficient. For the OS of areas larger (smaller) than 5000 km², the match is valid when the correlation of the pattern is greater than 0.6 (0.8). The correlation threshold values are consistent with those in Daniels et al. (2020). Otherwise, the OS BT₁₁ structures would change rapidly in one hour and would rather be considered stationary.

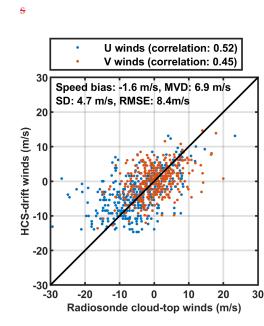


Figure 3. Comparisons of the U and V wind speeds between the tracked motions and radiosonde cloud-top winds within 150km at Darwin, Manus and Nauru in 2006 with a total sample number of 380.

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(3) Tracking OSs via dynamic overlaps

In Fig. 1b, to track the HCS temporal associationsOS temporal evolution of OSs, the OS is moved to the location predicted by cross correlation and then overlaps with the OSs in the later image. In this way, the dynamic overlaps b the dynamic overlaps are used for tracking the OS, which can be used to tolerate the avoids the mistakes in tracking the fast-moving OS in tracking. For the OS with single the corescore structure, three indices of the dynamic overlapping ratio (DOR) are considered to determine the associations of two OSs at different times for the same object, including the DOR between cores, the DOR between OSs and the DOR between cores and OSs. The DOR between cores is the ratio of their overlaps in cores relative to the minimum area of the cores to represent the degree of core overlap. The DOR between OSs is the ratio of the OS overlap relative to the minimum area of the OS to represent the degree of OS overlap. The DOR between cores and OSs is the ratio of the overlap of the OS to the core relative to the core area, representing the degree of the core overlapped by the later or previous

<u>OS.</u>

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Two OSs of different moments are associated in time and considered the same object when these two OSs overlap sufficiently. The overlapping situations of two OSs are distinguished by whether their cores overlap with each other (Fig. 1c) or not (Fig. 1d). Those pairs of OSs in situations (i), (ii) and (iv) in Fig. 1b all sufficiently overlap with the DOR between either 330 cores or OSs greater than 50% and thus are associated in time to reflect the OS evolution with time. The situation (iii) in Fig. Ic with DORs of both cores and OSs less than 50% indicates that these two OSs have no associations. In Fig. 1d, when the cores of two OSs do not overlap, the determinant of the OS association relies on the DOR between OSs and the DOR of OSs to cores. - and three indices of the core core, segmentation segmentation and core segmentation dynamic overlapping ratios (DORs) are used (Fig. 1b c). The core and segmentation DORs are relative to the minimum area and represent the degree of 335 the core and segmentation overlaps, respectively. The core segmentation DORs are relative to the core area and imply whether the core is inherited from the previous HCS. In Fig. 1b, four overlapping situations are distinguished by the core and segmentation DORs exceeding 50% or not. Three situations (i), (ii) and (iv) in Fig. 1b with DORs of either core or segmentation greater than 50% are recognized as the temporal associations, whereas the situation (iii) in Fig. 1b with DORs of both core and segmentation less than 50% disapproves the association of HCSs. In Fig. 1c, when the cores have no overlaps, the association 340 of HCSs relies on the DORs between segmentation and the DORs of segmentation to cores. In those cases, the OSs are associated in time only in situation (ii) in Fig. 1d, with those two DOR indices both larger than 50%, only situation (ii) in Fig. 1c can be recognized as the HCS association with those two DOR indices both larger than 50%, and tTheose pairs of HCSs OSs in the other situations in Fig. 1d of Fig. 1e are obviously not associated. Overall, if the DORs of two OSs satisfy the

overlapping conditions of (i), (ii) and (iv) in Fig. 1c and (ii) in Fig. 1d, they are associated in time. Thus, the conditions of (i), (ii) and (iv) in Fig. 1b and (ii) in Fig. 1c
 <u>(ii) and (iv) in Fig. 1b and (ii) in Fig. 1c</u>
 For two temporally associated OSs, the development and decay can be inferred from the variation in the cold-core BT₁₁

and area, as shown in Fig. 1e. The variation in the cold-core BT_{11} is prior to the variation in the cold-core areas for determining development and decay. If the cold-core BT_{11} is colder with time, or if the cold-core BT_{11} is the same but the cold-core area is larger with time, the OS is developing; otherwise, it decays. Notably, OSs are not necessarily associated with only one OS. As

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shown in Fig. 1e, mergers and splits are allowed in dynamic overlaps and are identified as the many-to-one and one-to-many OS associations, respectively. The merged cold-core BT_{11} and the splitting cold-core BT_{11} are documented as BT_{11} of mergers and splits, respectively.

In comparison, conventional fixed-threshold tracking infers development and decay from the variation in the area, which might involve multiple convective activities at different life stages. However, the variable-BT₁₁ tracking infers the development
and decay of each OS from the variation in both the cold-core BT₁₁ and the area. Additionally, mergers and splits in fixed-threshold tracking are dependent on the selection of the BT₁₁ threshold. Owing to the selection of the fixed BT₁₁ threshold, the identified targets are usually connected under a warmer threshold but are disconnected under a colder threshold. As illustrated in Fig. 1f, if under the fixed threshold of 260 K, no mergers or splits occur. If under the fixed threshold of 220 K, the cutoff of the CCO by 220 K is the connected complex of multiple cores or two disconnected parts at different times. This change in the connecting conditions over time under the selected fixed threshold results in mergers and splits in fixed-threshold tracking. If under the fixed threshold of 200 K, the mergers and splits of cold cores are captured. It manifests that mergers and splits in fixed-threshold tracking can be attributed to many reasons: the threshold selection, the change in the connecting conditions over time. In contrast, in variable-BT₁₁ tracking, mergers and splits are not influenced by changes in the connecting conditions over time but is only related to the variation in cold cores as illustrated in Fig. 1e-f.

365 (4) Quality control and validation of variable-BT₁₁ segment tracking

the HCS temporal associations are tracked by the conditions of (i), (ii) and (iv) in Fig. 1b and (ii) in Fig. 1e. Mergers and

splits are identified by many-to-one and one-to-many associations, respectively.

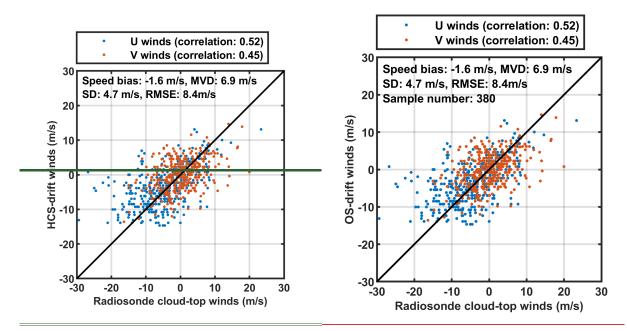


Figure 33. Comparisons of the U and V wind speeds between the tracked motions and radiosonde cloud-top winds within 150 km at Darwin, Manus and Nauru in 2006 with a total sample number of 380.

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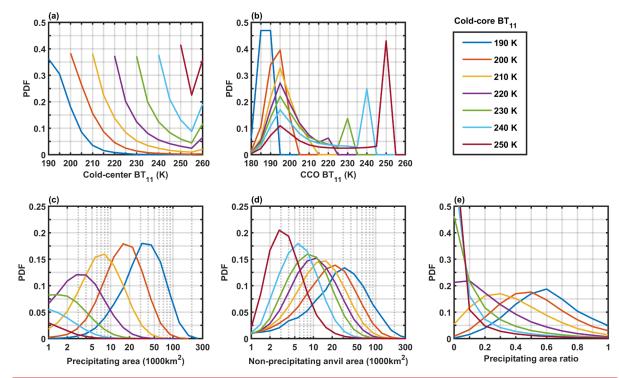
Quality control for missing images and the OSs touching the image edges is conducted. If the missing time gap between two continuous images exceeds 2 hours, the OSs in these two images and all lifecycles, including these OSs, are excluded from the analyses. Additionally, the OSs touching the image edges and all lifecycles, including the OSs touching edges, are excluded.

There are no direct observations to validate whether OSs are correctly associated. for the temporal association. ButHowever, some of the tracked behaviors (e.g., the tracked motions of OSs) can be compared against the observations measured by other senserssensors. Only if the tracking is correct, would the derived HCSOS-drift winds perform well. Thus, the OS-drift winds are compared against the radiosonde cloud-top winds at three ARM tropical sites in Darwin, Manus and Nauru. Here, the cloud-top winds are derived by combining the radar and radiosonde observations at those sites (see more details in Sect. 2.3) as the observational reference to examine the tracked OS motions from the hourly satellite images in 2006. To collocate the observations from the ground-based sites and satellites, the tracked OS-drift winds from the GEO observations that are closest to the time of the cloud-top wind observations and nearest to the site locations are compared with the cloud-top winds at those ground-based sites. The observational time difference is no more than one hour and the tracked OS core centroid is within 150 km of those ARM site locations. To match with the tracked motions of the organization segments, the radiosonde cloud top winds within 150 km of the segment centroids are considered as the observational reference, and the selection of this distance is These findings are consistent with those of previous studies in which examining the performance of cloud-trift winds is examined (Nieman et al., 1997; Santek et al., 2019; Daniels et al., 2020).

In Fig. 3-3, shows that the HCSOS-drift winds are significantly correlated with the high resolution-radiosonde cloud-top winds at the three long term ARM sites in Darwin, Manus and Nauru within 150 km. The correlations is are 0.52 and 0.45 at the 99% significance level for the U and V wind components, respectively, at the 99% significance level. On average, the HCSOS-drift winds are slower than the radiosonde-observed ambient-winds, with the a mean bias of -1.6 m/s. The A slow speed bias of 1-2-m/s is common in-for cloud-drift winds (Santek et al., 2019). Due-Owing to the limitations of the spatial and temporal resolutions (5-km and 1-hour, respectively), the least identifiable speed variation is approximately 5 km/hour (1.4 m/s), which is a possible reason for the slow speed bias. The bias in the mean angle bias is very small; approximately (0.5 degrees-), and tThe MVD, SD and RMSE are 6.9, 4.7 and 8.3 m/s, respectively. These biases are not surprising since the real-

world clouds do not strictly flow with ambient winds. In addition, some bias might be attributed to the uncertainty in the cloudtop heights. For its detection, the MMCR might underestimate the cloud top height since its signal would attenuate quickly for deep convective clouds (Hollars et al., 2004). In the convective systems, the motion of air is highly organized (Houze, 2004); thus, system movement might be inconsistent with the observed winds at the cloud-top height. ThTypically, the RMSE of the vector between the cloud-drift winds and the reference real-cloud-top winds is normally-approximately 6-13 m/s, according to previous studies (Santek et al., 2019; Bresky et al., 2012). This finding approves-indicates that the tracked motions of the OSsHCSs-are reasonable and thus the thereby -variable-BT₁₁ segment tracking is appropriate.

(5) Comparison with conventional fixed-threshold tracking



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Figure 4. The OS characteristics of CCOs. PDFs of cold-center BT_{11} (a), CCO BT_{11} (b), the precipitating area (c), the non-precipitating area (d) and the ratio of the precipitating area (e) for the OSs of the cold-core BT_{11} from 190-250 K.

The fundamental difference between fixed-threshold and variable-BT₁₁ tracking is target selection. With the fixed
 threshold of the BT₁₁, the connected convection of multiple cold cores is recognized as tracking targets, and only the area information is accessible. With the OS as tracking targets, variable-BT₁₁ tracking is capable of documenting the detailed evolution of each OS within CCOs, such as the developing depth, connecting conditions, and contributions to precipitation and anvil clouds.

The complexity of convective organizations determines the use of fixed-threshold or variable-BT₁₁ tracking. Specifically,
 if the structure of the convection is simple with only one cold core, it can be simply tracked via fixed-threshold tracking. Otherwise, for the complex organization of connected convective activities with multiple cold cores, variable-BT₁₁ tracking is suitable since it is capable of segmenting CCOs into OSs for tracking. The complexity of convective organizations can be inferred from the cold-center BT₁₁ of OSs. Only when the cold-center BT₁₁ is 260 K is the OS of the isolated convective body. Under the fixed BT₁₁ threshold, the OS of the cold-core BT₁₁ that is warmer than the selected threshold cannot be identified.
 The OS of the cold-center BT₁₁ that is colder than the selected threshold cannot be isolated from CCOs.

In Fig. 4, the OS characteristics (i.e., the connecting conditions with other surrounding OSs in CCOs and their contributions to precipitation and anvil cloud areas) of different developing depths with the cold-core BT_{11} from 190-250 K are investigated. In Fig. 4a, for the OSs of the cold core from 190-250 K, the probability distribution functions (PDFs) of the cold-center BT_{11} are shown. The PDF has a maximum peak of approximately 36-41% when the cold-center BT_{11} is equal to

- 425 the cold-core BT₁₁. For example, if in the fixed-BT₁₁ identification of the selected threshold of 220 K, only the OSs of the cold cores colder than 220 K and the cold centers warmer than 220 K can be isolated, but they account for only a small portion of the OSs of the cold-core BT₁₁ of 190-210 K. This implies that even under the cold BT₁₁ threshold, most of the identified targets still have complex organizations in the fixed-BT₁₁ identification. The warmer the selected BT₁₁ threshold is, the more complex the identified target is according to the fixed-threshold identification. The isolated structure with a cold-center BT₁₁ of 260 K
- 430 is rare for the deep convection of the cold-core BT₁₁ at 190-220 K, but it is relatively more frequent and seems to be another mode for the shallow warm systems of the cold-core BT₁₁ at 230-260 K. However, fixed-threshold tracking is not capable of distinguishing which target is the complex or isolated structure. In Fig. 2a, according to the probability distribution functions (PDFs) of the cold center BT_{H_2} it is the most frequent about 35 45% that the cold center BT_{H_2} is the same as the cold core BT_{II} and the isolated deep convective body in the 260 K shield is rare. The isolated structure is less frequent but seems to be
- 435 another mode for shallow warm systems.

Fig. 24b shows the PDFs of the complex CCO BT_{11} that, which refers to the coldest cold-core BT_{11} in the connected complex CCO. The CCO BT₁₁ can help to further distinguish the connecting conditions of the OS at different depths of development. If the CCO BT₁₁ is colder than the OS cold-core BT₁₁, the OS is in the deeper CCO and connected with the colder OS. Otherwise, if the CCO BT₁₁ is equal to the OS cold-core BT₁₁, the OS is the isolated structure or connected with

- 440 the warmer OS. For the OSs of the cold-core BT₁₁ from 200-260 K, the PDFs of the CCO BT₁₁ all peak at 195 K. It implies that they are the most frequently clustered in the 195-K CCO. For the OSs of cold-core BT₁₁ from 230-260 K, another peak of the PDFs of their CCO BT₁₁ is at their cold-core BT₁₁. Fig. 4a also shows that these OSs of the cold-core BT₁₁ from 230-260 K are more likely to have their cold-center BT₁₁ occur at 260 K. This implies that warm-core structures are more likely to be isolated as than cold-core structures. As a result, deep convective activities are mostly accompanied by the clustered complex 445 organization, and variable-BT₁₁ segment tracking is more suitable than fixed-threshold tracking for documenting their
- behaviors. It can be seen that the HCSs of the cold cores of 200 230K are the most frequently clustered in the 195 K complex, and the HCSs of the cores of 220 250K have two modes for the clustered 195 K complex and the isolated shallow warm structures, respectively. This implies the clustered complex organizations are the major mode for the deep convection and thus the segment tracking is necessary to distinguish their behaviors in the complex organizations.
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In variable-BT₁₁ tracking, precipitation and anvil clouds can be explicitly associated with unique cold cores. In Figs. 4cd, the distributions of the By segmentation, Figs. 2e e show the precipitation and anvil areas are lognormal and closely related to those of cold-core BT_{11} . The colder the cold-core BT_{11} is, the larger the precipitation and anvil areas to which the OS contributes. In Fig. 4e, the ratio of the OS precipitation area to the whole OS area is inversely proportional to the cold-core BT₁₁. Thus, the OSs of colder cores are dominated by precipitation, but these OSs still contribute to more anvil clouds than the 455 OSs of warmer cores. Similar to Lindzen et al. (2001), the ratio of the OS precipitation area to the whole OS area can also be understood as diagnostic of the precipitation efficiency or detrainment effect. The results in Figs. 4c-e might imply that the OSs of colder cores have increased precipitation efficiency, which contributes to both more precipitation and anvil clouds.

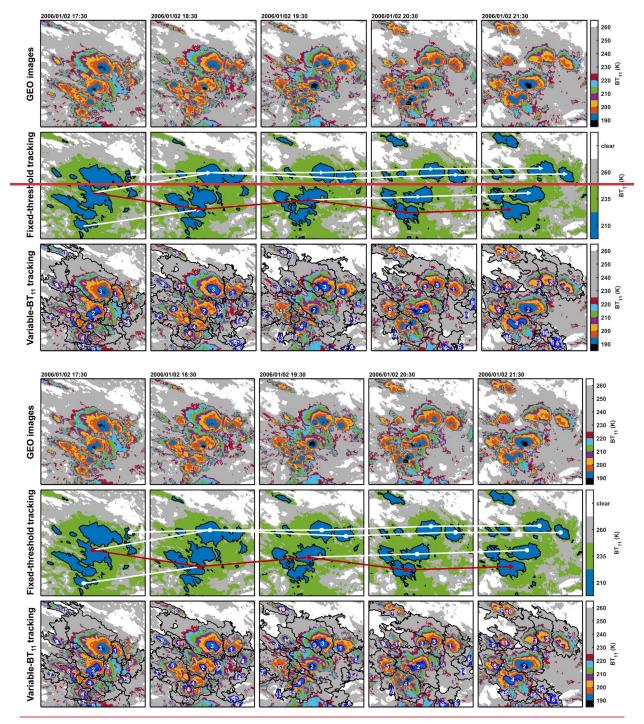
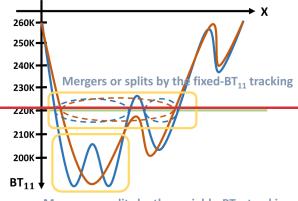


Figure 45. Examples illustrating the difference between the conventional fixed-threshold tracking -and the novel variable-BT₁₁ segment tracking. Uppermost panel: GEO BT₁₁ images taken between 5°S-15°S and 120°E-130°E from January 2, 17:30 to 21:30 UTC in 2006. Middle panel: the tracked lifecycles based on the fixed thresholds of 210, 235 and 260 K. The white lines represent the tree of the tracked temporal associationslifecycle by under the fixed threshold of 210 K and the red lines represent the main-major branch obtained by selecting the largest areas. Bottom panels: the variable-BT₁₁ identification and segment tracking. In the bottom panels, the blue contours indicate the cold cores, the black contours are the organization segmentsOSs in the the elustered convection complexCCOs, and the number at the core centroids indicates the lifecycle identification.

Examples of the conventional fixed-threshold and novel variable-BT₁₁ tracking algorithms are shown in Fig. 4-5 to
illustrate their differences. From the the GEO images in the uppermost panel of Fig. 45, it can be visually observed that, those clustered convections are connected in CCOs but have distinct behaviors in the complex to decay, split, develop and merge with over time. In the middle panel of Fig. 5, Under the fixed threshold of either 210, 235 or 260K, the convections and these ir behaviors of connected convections cannot be be are barely distinguished by fixed-threshold tracking from the complex in the middle panel of Fig. 4. With thresholds of 235 K and 260 K, the whole complex organization of connected convections

475 is identified as the tracking target. The whole clustered complex organization would be identified as one tracking target by the warm BT₁₁-thresholds. ByBy the cold threshold of 210 K, only a small -cold-part of the cluster complexCCO is identified; and nevertheless, those the connected convections of distinct behaviors are still cannot poorly be separated distinguished. Additionally, in fixed-threshold tracking at 210 K, -but the mergerss and splitss -are caused by variations in whether the convections are connected or disconnected under the 210-K threshold. due to the selection of the BT₁₁ threshold. In this case, 480 the tree of the tracked temporal associationslifecycle tree isare too complicated to analyze and is usually simplified by only selecting focusing only on the largest area at different times in each frame as the majorin lifecycle branch of the lifecycle. The main-major branch (the red line in the middle panel of Fig. 45) begins with the large complex organization of elustered connected convections complex that is becoming disconnected with time, andbut thus ends by with only one of disconnected partsless disconnected convection complex. Only the area information is available to describe the lifecycle in fixed-threshold 485 tracking. In contrast comparison, in the bottom panel of Fig. 5, with on the basis of the adaptive variable-BT₁₁ identification and segment tracking (the bottom panel of Fig. 4), it is shown that these connected systems convective activities are well separatedtely tracked as theinto decaying and splitting No. 3 HCSsOS, developing No. 2 HCSsOS, merging and developing No. 4 HCSsOS, etc. The mergers and splits of OSs are well tracked and not influenced by the variations in the connecting conditions over time. The area and cold-core BT₁₁ information are both available to describe the tracked lifecycles. And as 490 illustrated in Fig. 5 the mergers and splits are caused by the evolution of the system structures but not the variations of the connections.

The main difference between the fixed threshold and variable BT_{11} segment tracking is the selection of the tracking target. For the fixed threshold tracking, the tracking target is the 2-D area of the clustered convection complex. For the variable BT_{11} segment tracking, the tracking target is the 3-D structure of organization segments. The sizes and BT_{11} of the cold core represent the developing strength and the BT_{11} of the cold center represents the connecting conditions, and the segmentation distinguishes the area contribution of the convection in the clustered complex organization.



Mergers or splits by the variable-BT₁₁ tracking

Figure 5. Illustration of the difference for the mergers and splits captured by the variable BT_{μ} tracking and the fixed threshold tracking. The red and blue lines represent the evolution of the BT_{μ} structures of the complex organizations.

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4. Contributions of the the convective organization segments OS lifecycles to the precipitation and anvil cloud amount

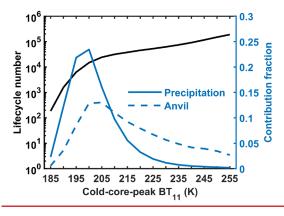


Figure 6. Sample numbers of tracked OS lifecycles with cold-core-peak BT₁₁ values from 185-255 K in the tropical western Pacific (130°W-170°E, 20°S-20°N) in 2006. The contribution fraction of the OS lifecycles to the precipitation and anvil cloud amount is shown on the right axis.

The warm pool of the tropical western Pacific Ocean (130°W-170°E, 20°S-20°N) is a typical region of oceanic convection precipitating and producing anvil clouds (Wall et al., 2018). In this section, only the OS lifecycles over the oceans in this region are considered for investigating the behaviors of the oceanic convection precipitating and producing anvil clouds. On the basis of the tracked lifecycles, the total precipitation and anvil cloud amounts can be attributed to two factors: (1) the lifecycle occurrence frequency and (2) the convection precipitating and producing anvil clouds during each lifecycle. With the novel tracking dataset, these two aspects are investigated in this section.

For OS lifecycles, the cold-core-peak BT₁₁ is identified as the coldest cold-core BT₁₁ in lifecycles and is used to represent the convective peaking strength. The OS lifecycles can be classified by the cold-core-peak BT₁₁. In Fig. 6, the sample numbers of lifecycles of different cold-core-peak BT₁₁ values over the tropical western Pacific Ocean in 2006 are shown. The warmer the cold-core-peak BT₁₁ is, the greater the number of lifecycles is. The lifecycle of the cold-core-peak BT₁₁ at 185 K has over one hundred samples, and the lifecycle of the cold-core-peak BT₁₁ at 255 K has hundreds of thousands of samples in one year. In addition, Fig. 6 also shows the fractions of the contributions of OS lifecycles of different cold-core-peak BT₁₁ values to the total precipitation and anvil cloud amounts. Here, the fraction of contribution refers to the sum of the precipitation (anvil) produced by all OS lifecycles in each bin of the cold-core-peak BT₁₁ divided by the total precipitation (anvil). The OS lifecycles of the cold-core-peak BT₁₁ values warmer than 235 K include a great number of samples, those lifecycles contribute to only no more than 5% of the total precipitation and anvil cloud amounts. Nevertheless, those warm OS lifecycles seem to be more important for anvil clouds than for precipitation.

(a 12 Non-precipitat<mark>i</mark>ng anvil (10⁵km²) 8 Precipitation (10¹⁰m³) Durat on (hour) 0.1 0.01 0.1 0.001 2 0.01 0.0001 10⁻⁵ 0.001 185 195 205 215 225 235 245 255 185 195 205 215 225 235 245 255 185 195 205 215 225 235 245 255 Cold-core-peak BT₁₁ (K) Cold-core-peak BT₁₁ (K) Cold-core-peak BT₁₁ (K)

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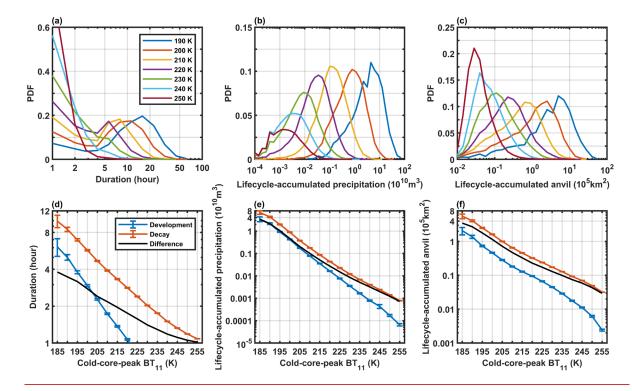


Figure 67. PDFs of the accumulated duration (a), precipitation (b) and non-precipitating anvil amount (c) of the OS lifecycles of different cold-core-peak BT₁₁ values from 190-250 K. The mean accumulated duration (d), precipitation (e) and non-precipitating anvil amount (f) contributed by the development (blue lines) and decay stages (red lines) as a function of the cold-core-peak BT₁₁ from 185-255 K. The black lines represent the differences in the accumulated duration, precipitation and anvil between the development and decay stages in (d-f), respectively. The error bars indicate the 95% confidence intervals of the means based on the t test.

The mean duration (a), precipitation (b) and non-precipitating anvil amount (c) contributed by the organization segments in development (blue lines) and dissipation stages (red lines) with the cold core peaking from 185-255K. respectively. The error bars indicate the 95% confidence intervals of the means based on the t test.

(Wall et al., 2018)By tracking the organization segments, the detailed convective activities and their contributions to the precipitation and non-precipitating anvil areas can be distinguished from the complex organization.

540 In Figs. 7a-c, for different cold-core-peak BT₁₁ values, the PDFs of the OS lifecycle-accumulated duration, precipitation and anvil amount are shown. In Fig. 7a, the PDFs of the duration of the cold-core-peak BT₁₁ that is warmer than 220 K peaks at 1 hour and most of them are less than 5 hours, whereas the lifecycle of the cold-core-peak BT₁₁ colder than 220 K has two modes for the short duration with the PDF peak at 1 hour and the long duration with the PDF peak at more than 5 hours. In Figs. 7b-c, the PDFs of the accumulated precipitation and anvil amounts contributed by the OS lifecycles basically conform to the lognormal distribution. Overall, the duration, precipitation and anvil amount of lifecycle accumulation are inversely proportional to the cold-core-peak BT₁₁.

In Figs. 67d-f, the relationships of the OS lifecycle-accumulated duration, precipitation and anvil amount with the cold-core-peak BT₁₁ are further investigated. The OS lifecycle is separated into development and decay stages. The development (decay) stage is defined as the stage before (after) the cold core peaks at the coldest BT₁₁ with the largest core area. In Fig. 7d, only the OS lifecycles of the cold-core-peak BT₁₁ colder than 220 K have a development of more than 1 hour, whereas the OS lifecycles of the peak BT₁₁ that is warmer than 220 K directly decay and disappear rapidly within a few hours. Figs. 7d-f show that the process of decay is longer than the process of development. Overall, the OS lifecycle-accumulated duration, precipitation and anvil amount have simple loglinear relationships with the peak BT₁₁ in both the development and decay stages. the development and dissipation stages of the tracked organization segments are separated by the time of the cold core peaking at the coldest BT₁₁ with the largest area. It is interesting that the convective behaviors, i.e., duration, precipitation and producing

the non-precipitating anvil clouds, all have a simple log-linear relationship with the peaking BT₁₁-in both the development and dissipation stages. According to the composites of the tracked lifetime (Fig. 6a), in the complex organizations, the segments of the cores colder than 220K would be more robust with the duration of 4-16 hours, while the shallow structures of the warmer cores are fragile and disappear rapidly. Figs. 6b c also show that the precipitation and anvil clouds in those cold structures are dozens fold as much as that of the warm structures. It is also noted that more precipitation and anvils are contributed by the dissipation stage of the organization segments and tThe difference of-in the accumulated duration, precipitation and anvils anvil cloud amounts between the two stages (the black line in Figs. 7d-f) also has exponentially increases, with the core peaking at colder BT₁₁ values. It seems that tThe convection duration of the development and decay processes –and the two key components of the convective its –cloud water budget, i.e., the lifecycle-accumulated precipitation and anvilsanvil cloud amounts, are simply determined byclosely related to its the peaking cold-core organization structures.



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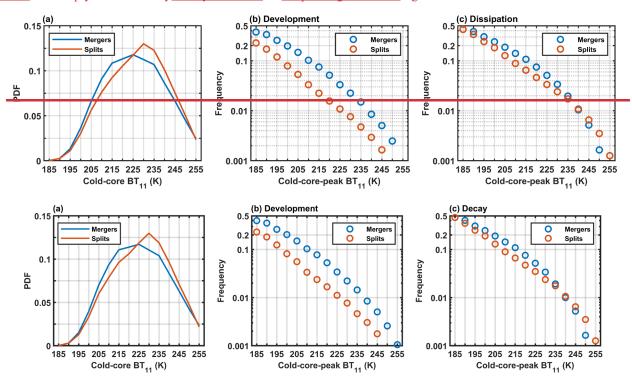
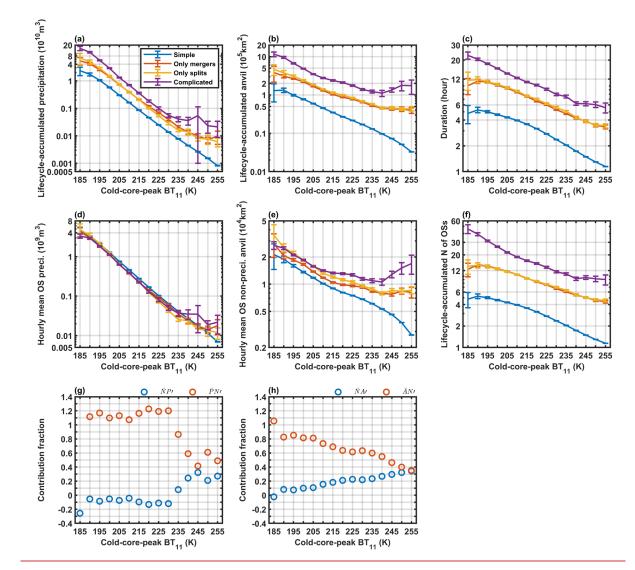


Figure 78. Observed mergers and splits of the convective structures. (a) The PDFs of of BT₁₁ values for mergers (blue lines) and splits (red lines) the merging (the blue line) and splitting (the red line) cold core BT₁₁. The frequency of the merging and splittingmerger and split occurrence in the evolution of the convection peaking lifecycles of the cold-core-peak BT₁₁ from 185K to _255 K in the development (b) and dissipation decay (c) stages, respectively.



575 Figure 9. (a-c) Composites of the lifecycle-accumulated precipitation, anvil cloud amounts and duration of different lifecycle types in each bin of the cold-core-peak BT₁₁ from 185-255 K, respectively. (d-e) Composites of the hourly mean OS precipitation and anvil cloud amounts, and the lifecycle-accumulated number of OSs of different lifecycle types in each bin of the cold-core-peak BT₁₁, respectively. The blue, red, yellow and purple lines indicate the simple, only-merger, only-split and complicated lifecycles, respectively. (g) The fractions of contributions of the OS hourly precipitation and the lifecycle-accumulated precipitation. (h) The fractions of contributions of the OS hourly anvil production and the lifecycle-accumulated OS number of anomalies to the variation in lifecycle-accumulated OS number of anomalies to the variation in the lifecycle-accumulated of the means based on the t test.

In the conventional fixed threshold tracking, mergers and splits mostly result from the connection or disconnection of the tracked clustered convective systems. By tracking the segmentation in the clustered complex, the mergers and splits reflect the detailed inner variations of the complex organizations, and represent more HCSs are born from the segmented convection and the complex organizations are clustered by more HCSs, e.g., the No. 4 segment for mergers and the No. 3 segment for splits in the bottom panel of Fig. 4 and the illustration in Fig. 5.

Fig. 7 shows that the mergers and splits are strongly determined by the BT₁₁. On the system level<u>In Fig. 8a</u>, according to
the PDFs of the cold-core BT₁₁ of mergers and splits, the cold-core BT₁₁ of mergers is more distributed at colder BT₁₁ values than that of splits. This implies that mergers are more likely to occur for the OSs of cold BT₁₁ values, whereas splits are more likely to occur for the OSs of warm BT₁₁ values. In Figs. 8b-c, according to the frequency of merger and split occurrence in the OS lifecycles of different cold-core-peak BT₁₁ values, in both the development and decay stages, it is somewhat surprising that the frequency of mergers and splits still has a loglinear relationship with the cold-core-peak BT₁₁. In the development process, mergers are more likely to occur than splits. In the decay process, mergers are relatively more frequent for

the cores colder than 225 K and less frequent for the warmer cores as compared with the splits (Fig. 7a). On the lifecycle level, mergers are more frequent than the splits in the development stage (Fig. 7b), and the splits in the dissipation stage (Fig. 7c) are more frequent than that in the development stages. In both the development and dissipation stages, it is a bit surprising that the frequency of the mergers and splits still has a log-linear relationship to the cold-core peak BT_H-

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According to the occurrence of mergers and splits, OS lifecycles can be further classified into simple (no mergers or splits), only-merger, only-split and complicated (both mergers and splits) types. In Figs. 9a-c, How does the mergers and splits influence the precipitation and anvil amount? On the lifecycle level, there would be two possible reasons: the merger and splits 605 influence the precipitation and anvil in the HCSs, or increase the HCS numbers. In Fig. 8, the lifecycles are classified into simple (no mergers and splits), only merger, only split and complicated (both mergers and splits) types. Figs. 8a c show that the lifecycle-accumulated total-precipitation-and, non-precipitating anvil cloud amounts and duration -area contributed by the tracked convection and its lifetime are strongly related to the occurrence of the mergers and splits. For the same cold-corepeak BT₁₁, the complicated lifecycles have the largest accumulated precipitation, anvil and duration among all lifecycle types. 610 The only-merger and only-split lifecycles have similar accumulated precipitation, anvil and duration values that are greater than those of the simple lifecycles. Interestingly, for different types of lifecycles, the slopes of the loglinear relationships of the lifecycle-accumulated precipitation, anvil and duration, with the cold-core-peak BT₁₁, are nearly invariant. This implies that mergers and splits do not influence the dependence of the lifecycle-accumulated precipitation and anvil cloud amounts on the BT₁₁ structures and that the increased precipitation and anvil cloud amounts caused by the mergers and splits also conform 615 to a loglinear relationship with the cold-core-peak BT₁₁.

How do mergers and splits influence the lifecycle-accumulated precipitation and anvil cloud amounts? There are two possible mechanisms: the hourly precipitation and anvil production of each OS in the lifecycle are enhanced, and the accumulated number (N) of OSs in the lifecycle is increased. In Figs. 9d-f, the hourly mean precipitation and anvil amount of each OS and N of OSs are shown for different lifecycle types. For the same cold-core-peak BT₁₁, the hourly mean OS precipitation of different lifecycle types (Fig. 9d) is nearly invariant. The hourly mean anvil production of OSs (Fig. 9e) is greater in the lifecycles with the occurrence of mergers and splits than in the simple lifecycles. In Fig. 9f, the accumulated N of OSs has the most significant variation for different lifecycle types. For the same cold-core-peak BT₁₁, N in the complicated lifecycles is much larger than the N of the other lifecycles. For the only-merger and only-split lifecycles, N is nearly the same and much greater than the N of the simple lifecycles. Thus, for different lifecycle types of mergers and splits, the variation in lifecycle-accumulated anvil amount is influenced by both the hourly anvil production in each OS and the N of OSs.

Accordingly, in Figs. 9g-h, the anomalies of the lifecycle-accumulated precipitation and anvil cloud amounts can be decomposed as follows:

$$PN - \bar{P}\bar{N} = \bar{N}P' + \bar{P}N' + P'N', \tag{7}$$

$$AN - \bar{A}\bar{N} = \bar{N}A' + \bar{A}N' + A'N'.$$
(8)

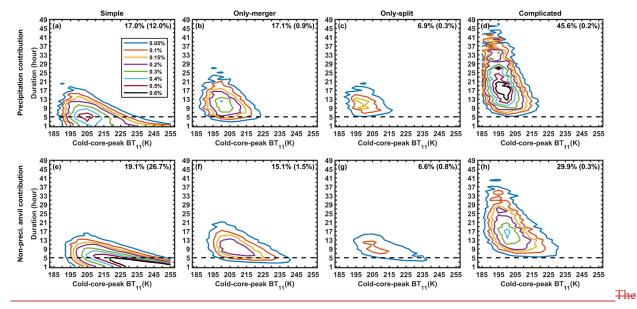
P and A are the OS hourly precipitation and anvil cloud amount, respectively. N is the accumulated OS number in the lifecycle. Thus, PN and AN represent the lifecycle-accumulated precipitation and anvil cloud amount, respectively. The bar over the letter represents the mean of different lifecycle types, and the prime represents the anomaly due to the different lifecycle types. In this way, NP' and PN' indicate the contributions of the OS hourly precipitation anomaly and the OS number anomaly, respectively, to the variation in lifecycle-accumulated precipitation. Similarly, NA' and AN' indicate the contributions of the OS hourly precipitation in the lifecycle-accumulated anvil amount. P'N' and A'N' are high-order small quantities and are neglected. The fraction of the contributions of NP' and PN' (NA' and the left-hand-side quantities of Eq. 7 and Eq. 8. Figs. 9g-h show the fractions of the contributions of NP' and PN' (NA' and PN')

 $\bar{A}N'$) to the increase in the lifecycle-accumulated precipitation (anvil) from the simple to complicated lifecycles. For the longlived lifecycles of cold-core-peak BT₁₁ values colder than 220 K, $\bar{N}P'$ has a small negative contribution of approximately -10%, whereas $\bar{P}N'$ has a large contribution of approximately 110%. And $\bar{N}A'$ and $\bar{A}N'$ both have positive fractions of contribution of approximately 0-20% and 65-105%, respectively. For the short-lived warmer lifecycles, although the contributions from $\bar{N}P'$ and $\bar{N}A'$ increase, the N anomaly still makes the greatest contribution to the variation in the lifecycleaccumulated precipitation and anvil cloud amounts. Overall, in comparison with simple lifecycles, the mergers and splits can create more OSs whose precipitation is slightly reduced and anvil production is enhanced, to ultimately prolong lifetime and increase both precipitation and anvil clouds of the lifecycle accumulation.

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more complicated the lifecycle is, the larger the precipitation, anvil and lifetime are. Mergers and splits seem to have the same impact on the precipitation and anvil, without significant differences for the lifecycles of only mergers and only splits. Figs. 8d f decompose the total precipitation and anvil into the contribution from each HCS and the number of the HCSs, respectively, as follows:

$DN = \overline{D}\overline{N} = \overline{N}D^{+} \pm \overline{D}N^{+} \pm D^{\prime}N^{\prime}$	(7)
$AN - \bar{A}\bar{N} = \bar{N}A^{t} + \bar{A}N^{t} + A'N'.$	(8)

P, A and N are the precipitation, non-precipitating anvil areas in each HCS and the HCS numbers, respectively. The bar over
the head and prime represent the mean and anomaly, respectively. It is shown that the HCS precipitation is nearly the same no matter whether mergers or splits occur (Fig. 8d) and contributes to -10% (22%) of the variations of the total lifetime precipitation for the peaking core colder (warmer) than 230K (Fig. 8g). In Fig. 8e, some variations of the HCS anvil amount can be found with the occurrence of the mergers and splits but are still relatively small with the mean contribution fraction of 19% to the variation of the total lifetime anvil amount (Fig. 8h). In Fig. 8f h, on average, the mergers and splits can significantly increase the HCS numbers to contribute to 98% and 67% of the variations of the total lifetime precipitation and

anvil amount, respectively. It implies that the mergers and splits could slightly impact the efficiency of precipitating and

producing the anvil in each HCS but mainly enhance the lifetime total precipitation and anvil amount by increasing the HCS numbers.

It is also interesting to note that the slope of the log linear relationship of the precipitation, anvil amount and lifetime to the cold core peak BT₁₁ is nearly invariant for different types of the lifecycles. It would mean that the mergers and splits do not influence the dependence of the precipitation and anvil on the organization structures and the increased precipitation and anvil by the mergers and splits have to conform to the log linear relationship to the cold core peak BT₁₁.

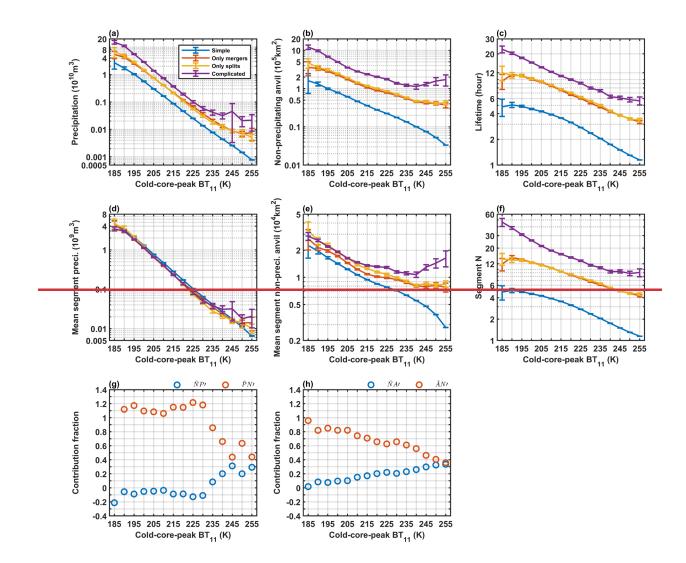


 Figure 8. Mean precipitation and anvil amount produced from the whole lifecycle and each single segment, respectively. (ac) The composites of the precipitation, non-precipitating anvil and lifetime contributed by the whole lifecycles, respectively. (d e) The composites of <u>The blue, red, yellow and purple lines indicate the simple, only merger, only split and complicated</u> <u>lifecycles.precipitation and non-precipitating anvil produced by each segment, respectively. (f) The total number of segments in the tracked lifecycles. (g) The contribution fraction of the segment precipitation and number for the variation of the total lifetime precipitation. (h) The contribution fraction of the segment anvil clouds and number for the variation of the total lifetime anvil amount. The blue, red, yellow and purple lines indicate the simple, only-merger, only-split and complicated lifetime anvil amount. The blue, red, yellow and purple lines indicate the simple, only-merger, only-split and complicated lifetime anvil amount. The blue, red, yellow and purple lines indicate the simple, only-merger, only-split and complicated lifetime anvil amount. The blue, red, yellow and purple lines indicate the simple, only-merger, only-split and complicated lifetime anvil amount. The blue, red, yellow and purple lines indicate the simple, only-merger, only-split and complicated lifetime anvil amount. The blue precipitate the 95% confidence intervals of the means based on the t test.
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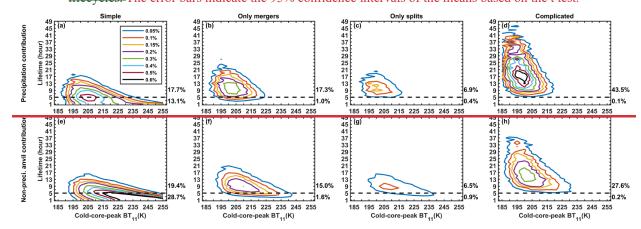


 Figure 910. Contribution Fractions of the contribution of the the tracked organization segmentsOS lifecycles of different coldcore-peak BT₁₁ values and durations to the total precipitation (a-d) and anvil cloud amount (e-h) over the tropical western Pacific Ocean for simple, only-merger, only-split and complicated lifecycles, respectively. (a d) Precipitation and (e h) nonprecipitating anvil contribution fraction of the segmented convection of different peaking strength and lifetime, for In each panel, the fractions of contributions of the long-lived (short-lived) lifecycles of durations larger (less) than 5 hours are listed in the top right-hand corner. simple, only-merger, only-split and complicated lifecycles, respectively.

685 InHow are the tropical precipitation and anvil associated with the convective organization? Here, the complex organizations have been segmented and tracked. It can be seen from Figs. 67-89, that the OS lifecycle-accumulated duration, precipitation and anvil cloud amounts of the tracked segment would increase exponentially have a loglinear relationship with the <u>cold-core-peak BT₁₁ colder BT₁₄</u> and are positively related to the occurrence of the mergers and splits, <u>However, cold</u> lifecycles are less common than warm lifecycles and thus might make only small contributions to the total precipitation and 690 anvil clouds (Fig. 6). In Fig. 10, the contributions of the OS lifecycles of different cold-core-peak BT11 values and durations to the total precipitation and anvil cloud amounts are shown. For precipitation, as shown in Figs. 10a-d, the complicated lifecycles have the largest contribution of 45.8%, compared with the simple (29%), only-merger (18%) and only-split (7.2%) lifecycles. The long-lived lifecycles of the duration at least 5 hours contribute to 86.6% of the total precipitation. For the nonprecipitating anvil clouds in Figs. 10e-h, the short-lived simple and the long-lived complicated organization segments lifecycles 695 and short-lived (less than 5 hours) simple lifecycles are two both-most important, with contribution fractions of 29.9% and 26.7%, modes to contribute to 28.7% and 27.6% of the non precipitating anvil clouds, respectively. The complicated longlived lifecycles, whose cold-core-peak BT11 is mostly less than 220 K, are efficient at producing anvil clouds but are less frequent. The short-lived simple lifecycles, whose cold-core-peak BT₁₁ are mostly greater than 220 K are not efficient at producing anvil clouds but are very frequent. On the whole, 68.5% of anvil are produced in the long lived organization segment, '00 including 19.4% from the simple lifecycles and 49.1% from the lifecycle with mergers and splits. A large portion of anvil (31.4%) is also attributed to the short lived warm organization segment. Overall, the precipitation is the most associated with the long lived cold organization segment of complicated lifecycles, while the non precipitating anvil is the most associated with both the short lived warm segment (frequent but not efficient to produce anvil) and long lived cold segments of complicated lifecycles (efficient to produce anvil but not frequent).but the warm organization segment (86% of the frequency 05 for the cold core peak $BT_{11} > 220K$) of the lifetime only a few hours are much more frequent than the cold organization segment (14% of the frequency for the cold core peak $BT_{H} \leq 220K$). Thus, Fig. 9 distinguishes the contribution of the organization segments to the total precipitation and anvil. In Fig. 9a d, 94% of precipitation are contributed by the organization segment of the cold core peak BT₁₁-colder than 220K. 85.4% of precipitation are produced from the long lived lifecycles (lifetime ≥ 5 hours). The segment of the complicated long lived lifecycle dominates the largest precipitation fraction of 43.5%, 10 as compared to the simple (30.8%), only merger (18.3%) and only split (7.3%) lifecycles. In Fig. 9e h, the short lived simple and long-lived complicated organization segments are two most important modes to contribute to 28.7% and 27.6% of the non-precipitating anyil clouds, respectively. On the whole, 68.5% of anyil are produced in the long lived organization segment, including 19.4% from the simple lifecycles and 49.1% from the lifecycle with mergers and splits. A large portion of anvil (31.4%) is also attributed to the short lived warm organization segment. Overall, the precipitation is the most associated with 15 the long lived cold organization segment of complicated lifecycles, while the non precipitating anvil is the most associated with both the short lived warm segment (frequent but not efficient to produce anvil) and long lived cold segments of complicated lifecycles (efficient to produce anvil but not frequent).

4.5. Conclusion

720 Tropical convection organizations are normally the connected complexes of a bunch of many convective activities. In this work, the clustered convection organizations are segmented and further tracked by a novel variable-BT₁₁ segment tracking algorithm is established to segment the CCO into OSs for tracking. The tracked motions of the segments OSs are compared against the observational winds for to examining examine the rationality of the tracking. Strong correlations between the tracked motions and real winds are found, with a-small differences inin the mean speeds (-1.6_m/s) and angles (0.5°). These results approve confirm that the tracking is appropriate.

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In comparison to Compared with the previous fixed-threshold and variable- BT_{11} tracking algorithms based on the variable

 BT_{11} or fixed- BT_{11} -identification, instead of that only focusing focus only on the 2-D area variation in the areas, the novel tracking algorithm developed in this work is capable of documenting the evolution of both the area and BT_{11} structure (i.e., the cold-core and cold-center BT_{11} for indicating the developing and connecting conditions, respectively, and mergers and

- 730 splits of cold cores). Precipitation and anvil clouds are explicitly associated with unique cold cores. The lifecycles are described by the variation in the cold-core BT₁₁, e.g., cold-core-peak BT₁₁, merger and split BT₁₁. The major difference between the fixed-threshold and novel variable-BT₁₁ tracking algorithms is the selection of tracking targets. The complex organizations of multiple cold cores are very frequent, and isolated convective bodies are rare, particularly for the deep convection of the cold-core.
- core BT₁₁ from 190-220 K. This implies that most of the targets identified by the fixed BT₁₁ threshold are complex
 organizations and their segmentation by the variable-BT₁₁ tracking algorithm is necessary. Additionally, in fixed-threshold tracking, the 3 D structures of the organization segments (i.e., cold core and cold center BT₁₁ for indicating the developing and connecting conditions, respectively, and the segmentation for distinguishing the area contribution of each convective activity in the complex) are identified and tracked. In the conventional fixed threshold tracking, the mergers and splits are easily-caused by many reasons: threshold selection, changes in the connecting conditions, and variations in the clod cores. This makes the final lifecycle in fixed-threshold tracking very complicated. In contrast, caused by the connectimergers and splits of cold cores in the variable-BT₁₁ tracking algorithm can be well captured and not influenced by the variation in the connecting

conditions. on or disconnection

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On the basis of the novel variable-BT₁₁ tracking algorithm, the convective processes of precipitation and the production of anvil clouds over the tropical western Pacific Ocean are investigated. due to the selection of the BT₁₁ threshold. But the mergers and splits of the structure body in terms of the same one object are tracked by the novel algorithm and reflect that the more HCSs are born from the convective activity and the cluster number of the complex organizations would increase. Overall, the detailed variations of the organization structure, especially on the BT₁₁-dimension, and the precipitation and anvil contributions of each organization segment can be tracked by the novel algorithm.

It is iInterestingly-to note that, the accumulated_duration, precipitation and anvil <u>cloud</u> amounts produced from the OS tracked organization segments lifecycles have a-simple log-linear relationships with the BT₁₁ of the cold core peak. Most of the lifecycles of the cold-core-peak BT₁₁ values colder than 220 K are long-lived, with durations of more than 5 hours. The lifecycles of the warmer-peak BT₁₁ are short-lived. During the lifecycle, decay requires more time than development does. More accumulated precipitation and anvil cloud amounts are also contributed by the decay stage than by the development stage. The difference in the accumulated duration, precipitation and anvil production between the two stages increases exponentially with decreasing cold-core-peak BT₁₁.

The organization segments of the cold core less than 220K are the most robust with the duration 4-16 hours, while the warmer organization segments are fragile to disappear rapidly but with the occurrence frequency as high as 86% in the complex. On the lifecycle level, more precipitation and anvil amount are contributed in the decaying stage of the organization segments as compared to that in the development stage. The difference between two stages would increase exponentially with the 60 decrease of the cold core peak $BT_{H^{+}}$. The occurrence of the mergers and splits also strongly depend-relies on the BT_{11} of the cold core peak, which still has a -with a log-linear relationship. Overall, mergers are more frequent than splits are in the development stage. In the decay stage, the frequency of mergers and splits shows little difference. The frequency of splits in the decay stage is greater than that in the development stage. Lifecycle-accumulated precipitation and anvil production are positively related to the occurrence of mergers and splits. With the same cold-core-peak BT11, the increase in the lifecycle-65 accumulated precipitation between the simple and complicated lifecycles is mostly attributed to the increase in the number of OSs during the lifecycle. The increase in the lifecycle-accumulated anvil cloud amount between the simple and complicated lifecycles should be attributed to the increase in both the hourly anvil production in each OS and the number of OSs. the mergers are more frequent at the colder cores in the development stages as compared to the splits, and the splits are more frequent in the decaying stages than that in the development stage. The organization segments accompanied with the mergers

- 70 and splits would produce more precipitation and anvil as compared to that with no mergers and splits. By the mergers and splits, more HCSs are born from the convection but on the lifecycle average the precipitation and anvil in each HCS only have small variations. The results show that for the complicated lifecycles 98% and 67% of the precipitation and anvil increases, respectively, are attributed to the increase of the HCS numbers. It is also interesting to note that tThe slope of the log-linear relationship between the lifetime-lifecycle-accumulated precipitation or anvil and the cold-core-peak BT₁₁ is almost invariant '75 in simple and complicated for different lifecycle typess.

Although the lifecycles of cold peaking cores are more efficient at precipitating and producing anvil clouds, it occurrence frequency is much lower than that of warm peaking cores. For the total cloud water budget of the tropical western Pacific Ocean, most of the total precipitation (45.6%) is contributed by long-lived complicated lifecycles, whereas short-lived simple lifecycles are less important for precipitation, with a contribution of 12%. However, for the anvil cloud amount, the long-lived

780 complicated lifecycle and the short-lived simple lifecycle are both important, with contributions of 29.9% and 26.7%, respectively, in the complex convection organizations, the organization segments of cold structures and long lived lifecycles with mergers or splits contribute to the largest fraction of the precipitation and non precipitating anvil amount, 67.7% and 49.1%, respectively. The warm short-lived organization segments contribute to little precipitation but relatively a high fraction of non-precipitating anvil (28.7%) due to its high frequency.

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Author contribution

ZW and JY designed the algorithm and, ZW carried out the experiment and prepared the manuscript.

790 Data and code availability

All data used in this study are available online. The GEO images (Nasa/Larc/Sd/Asdc, 2017) are obtained from the National Aeronautics and Space Administration (NASA) Langley Research Center Atmospheric Science Data Center (https://search.earthdata.nasa.gov/). GPM (Huffman, 2023) is obtained from the Goddard Earth Sciences Data and Information Services Center (GES DISC). The ground-based cloud (S. Giangrande, 1999) and wind (E. Keeler, 2001) observations at the ground-based sites are obtained from the Atmospheric Radiation Measurement user facility, a U.S. Department Of Energy (DOE) office of science user facility managed by the biological and environmental research program (https://www.arm.gov). The code of the anvil tracking algorithm is available upon request.

Competing interests

The author declares no conflict of interest.

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