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# Does discontinuous Mesoscale Convective System produce stronger precipitation?

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Mesoscale Convective Systems (MCSs) dominate global precipitation, yet their discontinuous evolution (merging, splitting, complex) and associated mechanisms remain unclear. Using Himawari-8 and IMERG datasets, we classify MCSs into four primary types (continuous, merging, splitting, complex), accounting for 97.15% of MCS-related precipitation. Discontinuous-type MCSs contribute 68.81% of total precipitation, demonstrating their critical role in freshwater supply, agriculture, and ecosystem stability, while posing greater flood risks to human societies. Spatial analyses reveal dual-core maxima over the ITCZ and East Asia. Discontinuous types exhibit consistent spatial patterns relative to continuous MCSs, despite intensity hierarchy (merging > complex > splitting). They enhance ITCZ rainfall by 4–16% via intensified updrafts and colder cloud tops, contrasting sharply with East Asia's precipitation suppression (–16% to 0) tied to disrupted cloud structure and reduced convective instability. The findings provide critical insights for improving flood early-warning systems and optimizing agricultural water management in monsoon-vulnerable regions, directly supporting climate adaptation strategies and sustainable development goals.

Mesoscale Convective Systems (MCSs) are primary contributors to precipitation in tropical and subtropical regions<sup>1</sup>. The heavy rainfall events they trigger often lead to floods, profoundly impacting agriculture, water resource management, and socioeconomic activities<sup>2</sup>. Globally, MCSs account for approximately 50% of total precipitation, exceeding 70% in extreme rainfall events<sup>3–5</sup>. Beyond modulating atmospheric circulation through latent heat release, MCSs also influence regional energy balance via cloud-radiation feedback, making them critical components in coupled weather-climate studies<sup>6</sup>. Therefore, a deeper understanding of MCS evolution and associated precipitation mechanisms is vital for improving extreme weather prediction capabilities and optimizing parameterization schemes in atmospheric models<sup>7</sup>.

The life cycle of MCSs—spanning initiation, development, maturity, and dissipation—is closely tied to precipitation efficiency. Classical studies have established statistical relationships between MCS area, duration, and rainfall intensity using radar and satellite observations<sup>8</sup>. Recent advancements in high-resolution satellite data have facilitated the development of MCS tracking techniques, enabling quantitative analyses of motion vectors, cloud-top temperature evolution, and life cycle characteristics<sup>9</sup>. Typically, MCSs produce the heaviest rainfall at around one-third of their life cycle <sup>10,11</sup>. However, existing research predominantly focuses on continuous-type

MCSs, leaving a gap in systematic exploration of discontinuous evolution processes (e.g., splitting, merging) and their dynamic-thermodynamic mechanisms.

Merging and splitting processes are critical dynamical drivers of MCS-related precipitation efficiency. Case studies revealed that merging processes generally enhance low-level convergence, prolong system longevity, and intensify rainfall<sup>12–14</sup>, while splitting processes have a more variable impact, potentially suppressing convection or triggering new systems<sup>15,16</sup>. Additionally, cloud-cloud interactions, such as microphysical adjustments from multi-system collisions, can significantly alter precipitation particle spectra<sup>17,18</sup>. Nevertheless, current studies rely heavily on case analyses or idealized simulations, lacking statistical characterization of real-world MCS evolution types, their regional dependencies, and quantitative linkages to precipitation fields.

Despite progress in MCS evolution research, key challenges remain in systematically evaluating the precipitation contributions and physical mechanisms of discontinuous MCS types and clarifying the regulatory pathways of MCS evolution on precipitation across diverse environments. Addressing these gaps, this study employs Himawari-8 satellite data and GPM precipitation products to develop a tracking-driven classification algorithm for MCS evolution, analyzing spatiotemporal patterns of variant evolution types. Through dynamical-statistical diagnostics, we reveal

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precipitation disparities in discontinuous MCSs across different regions, highlighting cross-scale interactions between environmental factors and intrinsic processes.

### Results

This study classifies MCSs into four primary categories and four secondary categories. The four primary MCS evolution types, including one continuous type and three discontinuous types (merging, splitting, and complex), account for 62.14% of the total MCS population but contribute 97.15% of the total MCS-related precipitation. In contrast, the four secondary MCS types (sub-continuous, sub-merging, sub-splitting, and dissipation) constitute 37.86% of the total MCS population yet contribute only 2.85% of the total precipitation. Given the significant disparity in precipitation contributions, this study focuses on the four primary evolution types of MCSs.

We first analyze the spatial distribution characteristics of precipitation associated with four types of MCSs. Observational results reveal that the mean precipitation distributions across all four MCS types align closely with the overall precipitation field (Fig. S2 in the Supplementary File), collectively displaying a dual-core spatial structure (demarcated by dashed rectangles in Fig. 1). The primary precipitation core is anchored over the Intertropical Convergence Zone (ITCZ), while the secondary core extends along the East Asian coastal belt (Hereafter East Asia). Both regions exhibit sustained precipitation intensities exceeding 2.5 mm h<sup>-1</sup> (Fig. 1a). Additional high-intensity precipitation centers (>2.5 mm h<sup>-1</sup>) are identified over the Bay of Bengal (BOB) and the southern Himalayan slopes. Notably, precipitation intensities in high-latitude and inland arid regions remain comparatively subdued, attributable to moisture

transport limitations and markedly reduced cloud-top heights. Nevertheless, most areas in these regions maintain baseline precipitation intensities above  $0.3~\mathrm{mm}~\mathrm{h}^{-1}$ .

Regarding the impact of discontinuous MCS processes on precipitation intensity, the initial theoretical hypothesis suggested that the splitting process of MCS accompanied by divergent flow fields might inhibit heavy precipitation development<sup>19</sup>, while merging processes associated with convergent fields should enhance precipitation<sup>14,20</sup>. However, quantitative analysis results contradict these initial assumptions. By calculating precipitation intensity difference fields between merging, splitting, and complex MCSs relative to continuous MCSs (Fig. 1b–d), the three discontinuous MCS types show highly consistent spatial patterns in precipitation differences, varying only in magnitude. This phenomenon suggests that discontinuous MCS types may share similar dynamic–thermodynamic precipitation mechanisms.

Specifically, within the ITCZ domain (5°S–15°N, 90°E–180°E), the discontinuous MCSs produce 0.2–0.8 mm  $h^{-1}$  (around 4–16%) stronger precipitation than continuous MCSs. Conversely, the East Asian coastal sector (30°N–45°N, 100°E–160°E) exhibits pronounced negative anomalies (–0.8 to 0 mm  $h^{-1}$ , around –16% to 0). A moderate positive anomaly (~0.2 mm  $h^{-1}$ ) characterizes the BOB, while the southern Himalayas persist as a precipitation stronghold dominated by continuous-type MCS activity. Besides, discontinuous MCSs display enhanced precipitation in moisture-limited environments, particularly across high-latitude zones and continental arid regions.

To rigorously assess the spatial consistency of precipitation difference fields among discontinuous MCS types, we employ statistical

Fig. 1 | Horizontal distributions of MCS-related rain rate. Average a rain rate produced by continuous-type MCSs and b-d rain rate differences between three discontinuous- and continuous-type MCSs during June–August 2016–2020. Zero-rain-rate grids are also incorporated in the statistics. The northern dashed rectangle indicates East Asia, while the southern one indicates ITCZ.

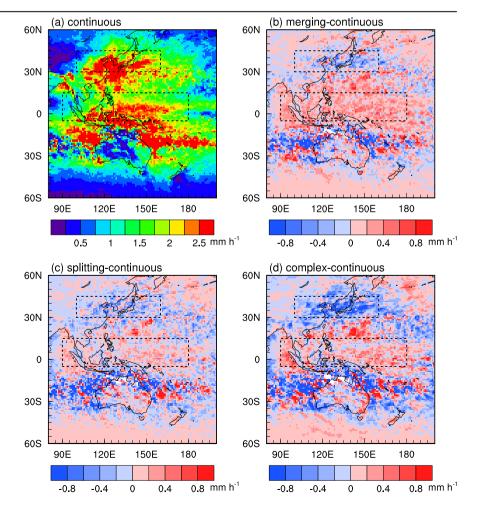
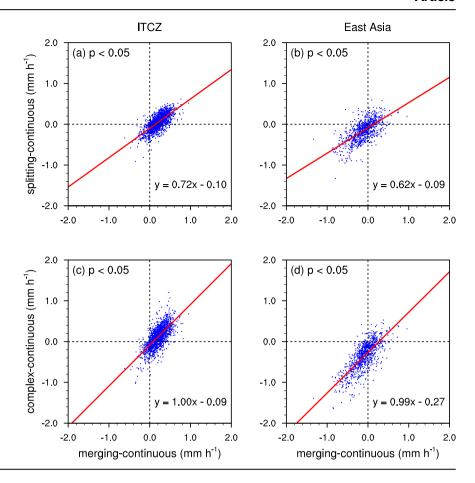


Fig. 2 | Scatter plots of average rain rate differences among discontinous MCS types. Scatters

**a**, **b** between merging-continuous and splitting-continuous and **c**, **d** between merging-continuous and complex-continuous over the ITCZ and East Asia. Each point represents a  $1^{\circ} \times 1^{\circ}$  grid in the study region, with the red line indicating the linear regression fit of scatters.



correlation analysis and regression-based analytical methods. Analyses are conducted at  $1^{\circ} \times 1^{\circ}$  resolution from Fig. 1b–d, incorporating only grid cells with >50 samples to ensure statistical reliability. Statistical validation demonstrates significant positive intercorrelations (p < 0.05) between precipitation anomaly fields of merging-, splitting-, and complex-type MCSs, mathematically verifying their shared spatial coupling patterns (Fig. 2).

In the regression analysis framework, precipitation anomalies from merging-type MCSs (calculated merging minus continuous) serve as the independent variable (x-axis). The regression equation established between rainfall anomalies of merging- and splitting-type MCSs (Fig. 2a, b) features a slope of 0.72 (0.62) and a negative intercept of -0.10 (-0.09) over ITCZ (East Asia). These parameters suggest that under equivalent geographical forcing, splitting-type MCSs demonstrate systematically reduced precipitation intensification capacity, exhibiting approximately 30% weaker enhancement compared to merging-type systems. In contrast, complex-type MCSs (Fig. 2c, d) follow the relationship y = 0.89x - 0.09 (y = 0.99x - 0.27) over ITCZ (East Asia), indicating nearly linear proportionality with merging-type anomalies. However, the negative intercept also reveals a persistent systematic deficit of 0.09 (0.27) mm h<sup>-1</sup> in precipitation intensity relative to merging-type counterparts.

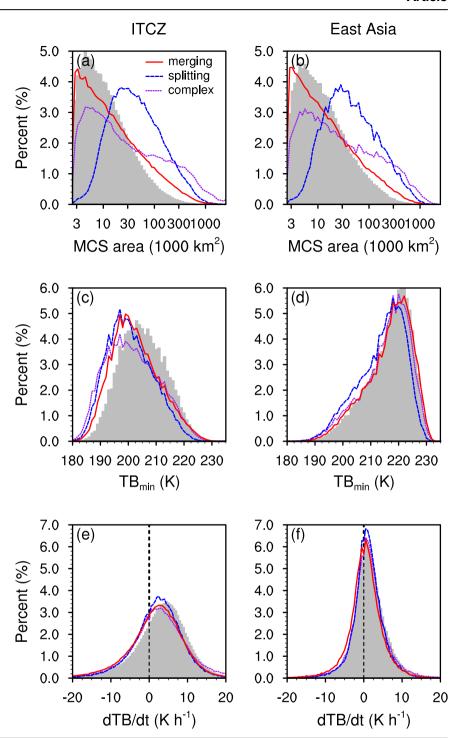
This gradient hierarchy also demonstrates the local precipitation intensity ranking: merging-type > complex-type > splitting-type. Combined with previous researches, this hierarchy may originate from (1) merging processes enhance low-level convergence through multi-cell interactions, promoting sustained updraft development; and (2) splitting processes accompanied by mid-upper-level divergent flows may prematurely consume convective available potential energy, limiting precipitation intensity.

To elucidate the intrinsic mechanisms governing precipitation differences between non-continuous and continuous MCS types, we conducted probability density function (PDF) analyses of key physical parameters of MCS across ITCZ and midlatitude East Asian coastal regions (Fig. 3), focusing on three diagnostic dimensions: horizontal scale, cloud-top thermodynamic state, and evolutionary processes.

The horizontal scale of MCS is positively related to precipitation intensity (Houze et al.8). Our analysis reveals analogous MCS area distributions across key regions. MCS area span three orders of magnitude  $(2.5 \times 10^3 - 1.0 \times 10^6 \text{ km}^2)$ , encompassing meso- $\beta$  ( $<3 \times 10^4 \text{ km}^2$ ) to meso- $\alpha$  (>3 × 10<sup>4</sup> km<sup>2</sup>) convective systems (Fig. 3a). Continuous-type MCS exhibit peak occurrence at  $\sim 4.5 \times 10^3$  km<sup>2</sup>, while merging-, splitting-, and complex-type MCS peak at  $3.1 \times 10^3$ ,  $2.22 \times 10^4$ , and  $4.5 \times 10^3$  km<sup>2</sup>, respectively. Meso-α MCS account for 18.0%, 27.5%, 58.6%, and 43.2% of continuous, merging, splitting, and complex types. This scaling hierarchy suggests that Splitting processes predominantly occur in larger-scale MCS systems, likely associated with dissipation-phase dynamical instabilities. Complex-type MCS in favor of intermediate scales reflects enhanced cloud-process interactions in extended systems. The striking spatial similarity in scale distributions between ITCZ and East Asia precludes horizontal scale as the primary driver of precipitation disparities in the key regions.

The minimum  $10.4 \,\mu\text{m}$  brightness temperature of MCS (TB<sub>min</sub>), a proxy for convective vigor<sup>21</sup>, exhibits distinct regional PDF patterns (Fig. 3c, d). The ITCZ displays dominant TB<sub>min</sub> peaks at 195–200 K (deep convection), contrasting with East Asian maxima near 220 K (weaker convection) consistent with monsoonal frontal precipitation characteristics<sup>22</sup>. Notably, discontinuous MCS systematically exhibit colder cloud tops than continuous types (ITCZ: 197 K vs. 203 K; East Asia: 220 K vs. 222 K), particularly demonstrating 6 K enhancements in ITCZ convective intensity.

Fig. 3 | Probability density functions (PDFs) of key physical parameters of MCS. PDFs of a, b MCS area, c, d minimum brightness temperature ( $TB_{min}$ ), and e, f temporal variations of  $TB_{min}$  (dTB/dt) over the ITCZ and East Asia. The gray shaded areas represent continuous-type MCSs, and colored lines denote discontinuous-type MCSs.



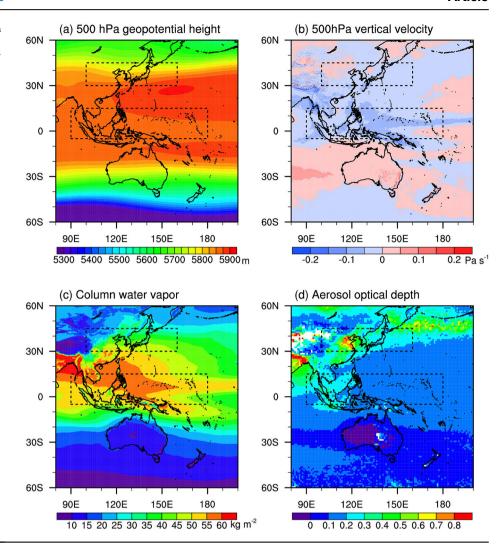
Splitting-type MCS show marginally taller tops than merging/complex types, ordered as: splitting > complex > merging > continuous.

The temporal variations of  $TB_{min}$ , quantified by its time derivative (dTB/dt), serve as a diagnostic metric for MCS life-cycle stages <sup>10</sup>: negative dTB/dt values (cloud-top cooling) indicate the intensification stage, near-zero dTB/dt values (thermodynamic equilibrium) correspond to the mature stage, and positive dTB/dt values (cloud-top warming) mark the dissipation stage. All MCS types exhibit quasi-normal dTB/dt distributions (Fig. 3e, f), with ITCZ systems showing greater variability  $(dTB/dt \in [-20, 20] \text{ K h}^{-1})$  than East Asian counterparts  $(dTB/dt \in [-10, 10] \text{ K h}^{-1})$ . The right-skewed dTB/dt peaks across regions reflect prolonged dissipation phases relative to intensification periods. Notably, discontinuous MCS in

the ITCZ exhibit  $2\,\mathrm{K}\,\mathrm{h}^{-1}$  stronger cooling rates than continuous types (peaks of PDFs), indicating enhanced convective triggering during discontinuous processes. Conversely, the PDFs of MCS types over East Asia show minimal inter-type divergence, suggesting regional environmental constraints dominate over intrinsic MCS variability.

The multivariate analysis of precipitation determinants, including moisture transport, dynamical forcing, and aerosol-cloud interactions, reveals key regional contrasts (Fig. 4). In the ITCZ region (Fig. 4a, c), the persistent low-level convergence driven by Hadley circulation induces deep ascent, with average vertical velocity reaching  $-0.07~{\rm Pa~s^{-1}}$  at 500 hPa. High column moisture levels of 50–60 kg m $^{-2}$  sustains moisture supply as well as latent heat release, while moderate aerosol conditions characterized by

Fig. 4 | Horizontal distributions of precipitation determinants. Average a 500 hPa geopotential height, b 500 hPa vertical velocity, c column water vapor, and d aerosol optical depth during boreal summer months (June–August) between 2016 and 2020.



aerosol optical depth (AOD) values of 0.1–0.3 minimize precipitation suppression.

In the East Asian region, mid-tropospheric flow is dominated by baroclinic westerlies, marked by a 500 hPa geopotential height gradient of 80 gpm per 10° latitude. The sharp meridional moisture, which declines from 50 kg m<sup>-2</sup> in the southern portion to 25 kg m<sup>-2</sup> in the northern portion, constrains northerly moisture supply in the East Asia. Heavy aerosol loading with AOD > 0.4 enhances cloud droplet competition, delaying warm rain initiation<sup>17</sup>. Weak mean ascent ( $\omega = -0.03 \, \mathrm{Pa \, s^{-1}}$  on average) disrupts the generation of deep convection.

To figure out why discontinuous processes exert opposite impacts over the ITCZ and East Asia, the thermodynamic parameters associated with different types of MCS events are further characterized (Fig. 5). In the ITCZ region, where atmospheric moisture content is exceptionally high (Fig. 4c), the development of MCSs do not rely on moisture transport by strong horizontal wind fields. The median vertical moisture flux divergence approaches zero (Fig. 5a), accompanied by relatively weak horizontal wind shear with a median value of approximately 8 m s<sup>-1</sup> (Fig. 5b). This MCSs have intense convective activity (Fig. 3c), as evidenced by a median CAPE value reaching 600 J kg<sup>-1</sup> (Fig. 5c). During discontinuous processes within MCSs, both moisture flux convergence and horizontal wind shear show moderate increases while CAPE remains essentially stable, creating favorable conditions for updraft (not shown) and precipitation enhancement.

In contrast, the East Asia region displays distinct characteristics due to its heterogeneous moisture distribution. Cloud and precipitation formation

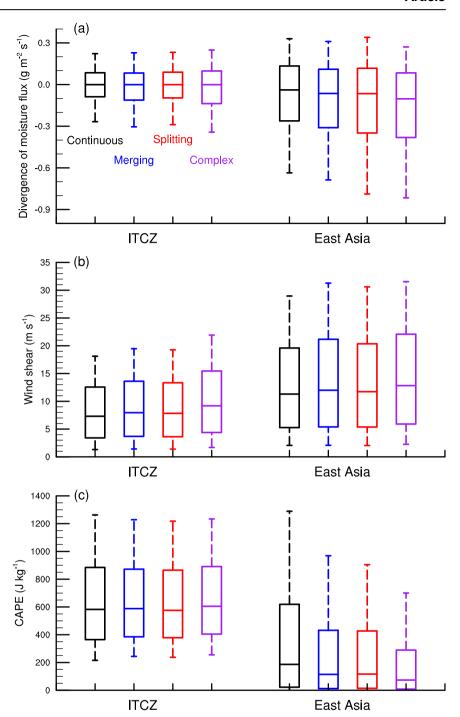
in this area critically depend on moisture transport through strong southwesterly winds (Fig. 4a, c). MCSs consequently demonstrate stronger moisture flux convergence and enhanced horizontal wind shear, with median values reaching approximately  $12~{\rm m~s^{-1}}$ . Dominated by stratiform precipitation (Fig. 3d), this region shows significantly lower convective potential with a median CAPE below 200 J kg $^{-1}$  (Fig. 5c). During discontinuous processes within MCS, while moisture flux convergence and wind shear exhibited increases, CAPE undergoes dramatic reduction from  $186~{\rm J~kg}^{-1}$  to approximately  $100~{\rm J~kg}^{-1}$ , corresponding to precipitation suppression mechanisms.

### **Discussion**

This study elucidates the pivotal role of discontinuous MCSs in shaping regional precipitation patterns through a synthesis of satellite observations and dynamical diagnostics. The primary types of MCSs, including continuous- and discontinuous-type (merging, splitting, and complex) MCSs, despite constituting 62.14% of total MCS systems, dominate MCS-related precipitation contributions (97.15%). With a larger area than continuous MCSs, discontinuous-type MCSs exhibit even more efficient contribution. Complex-type systems exhibit 33.13% rainfall contribution, followed by merging (21.15%) and splitting types (14.53%), highlighting their critical role in extreme rainfall events.

Despite hierarchical intensity differences (merging > complex > splitting), all discontinuous MCSs exhibit spatially coherent anomaly patterns relative to continuous systems, suggesting shared thermodynamic-dynamic frameworks. In the ITCZ, discontinuous

Fig. 5 | Variations of thermodynamic parameters among different types of MCS. a Divergence of vertically integrated moisture flux, **b** wind shear between 200 and 850 hPa, and **c** CAPE of different types of MCSs over ITCZ and East Asia. The box plot here indicates the 90%, 75%, 50%, 25%, and 10% percentiles of the variable.



systems enhance rainfall by  $0.2-0.8\,\mathrm{mm\,h^{-1}}$ , whereas they suppress precipitation by -0.8 to  $0\,\mathrm{mm\,h^{-1}}$  in East Asia. Through systematic analysis, we further reveal distinct variations in large-scale environmental conditions, MCS characteristics, and in-cloud thermodynamical parameters between these two regions. Building on these findings, a conceptual framework (Fig. 6) that elucidates how discontinuous processes differentially modulate precipitation mechanisms across the ITCZ and East Asia is proposed:

For the ITCZ region (Fig. 6a), within deep convective precipitation regimes characterized by homogeneous high moisture availability and optimal aerosol conditions<sup>23</sup>, discontinuous processes primarily amplify moist convective processes through intensified divergence of moisture flux and updraft though generating secondary circulations<sup>24</sup>, thereby enhancing precipitation efficiency.

In the East Asian monsoon region (Fig. 6b), MCS precipitation is predominantly governed by frontal processes, where convective contributions remain limited with less than 10% of convective ratio in Meiyu frontal systems<sup>25</sup>. It is suggested that discontinuous MCS processes exert precipitation suppression through disrupting the structure of stratiform clouds via strong wind shear, which subsequently reduces CAPE. Under moisture-deficient and aerosol-rich conditions (Fig. 4c, d), aerosol-mediated hydrometeor competition further suppresses warm rain formation<sup>26</sup>.

Critically, while maintaining the overarching mechanistic framework, intrinsic differences emerge among the three discontinuous MCS types. Merging processes enhance precipitation through cloud-scale convergence and collision-coalescence efficiency<sup>12</sup>. Splitting processes promote dry air entrainment and mid-level divergence, accelerating hydrometeor

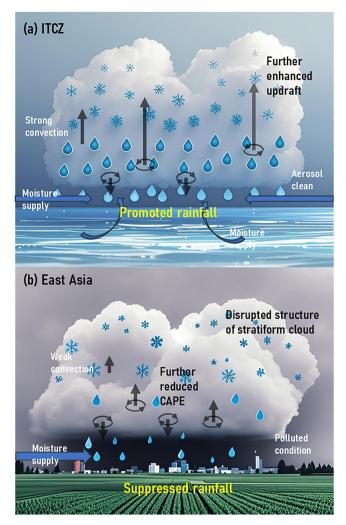


Fig. 6 | Concept of the opposite impact of discontinuous MCS evolution process. Impact of discontinuous process on precipitation intensity in a ITCZ and b East Asia. The elliptical arrows represent the secondary circulation induced by discontinuous MCS processes; the thick black arrows indicate updrafts and downdrafts; the blue arrows represent moisture transport; and the blue snowflake and raindrop symbols represent precipitation.

evaporation<sup>27</sup>. Complex processes, which dynamically integrate elements of both merging and splitting processes, demonstrate intermediate precipitation impacts, with their net effects contingent on the dominance of coalescence versus evaporation pathways under specific environmental constraints.

These findings advance mechanistic understanding of MCS evolution and emphasize the necessity of incorporating cross-scale interactions between intrinsic dynamics and environmental drivers into rainfall prediction. Such improvements are critical for enhancing flood prediction accuracy in vulnerable regions like the ITCZ and East Asia, where 70% of annual flood fatalities are linked to MCS-driven extremes. Future work should integrate high-resolution simulations to unravel microphysical causality and refine parameterizations for extreme precipitation forecasting, directly addressing the World Meteorological Organization's "Early Warnings for All" initiative aimed at protecting 3 billion people from climate disasters by 2030.

### Methods

## **Evolution categories of MCSs**

We utilize Level-1B observation data from the Advanced Himawari Imager (AHI) onboard the Japan Meteorological Agency's Himawari-8 geostationary

satellite, acquired via the JAXA P-TREE system (ftp://ftp.ptree.jaxa.jp). Data from boreal summer months (June–August) between 2016 and 2020 are selected, with a temporal resolution of 10 min and a spatial resolution of 0.05°  $\times$  0.05° ( $\sim$ 5 km). The 10.4  $\mu m$  band brightness temperature (TB $_{10.4}$ ) data is primarily employed, as this band is less affected by atmospheric scattering effects (atmospheric transmittance >85%), effectively characterizing cloud-top height and phase properties  $^{28}$ .

We identify MCS over the central coverage of Himawari-8 AHI (60°S–60°N, 80°E–160°W) following the classical definition by Mapes and Houze (1993) with two key criteria<sup>29</sup>:

- 1) Spatial continuity: an MCS is defined as a contiguous region where adjacent grids exhibit TB  $_{10.4} \le 235$  K.
- 2) Scale threshold: the effective coverage area of an MCS must be  $\geq 2.5 \times 10^3 \text{ km}^2$  (~100 grids).

Building upon the established MCS identification criteria, we develop a tracking-driven algorithm to systematically categorize MCS evolution types. The workflow is structured as follows:

Step 1—Temporal matching. For two consecutive observation times  $T_0$  and  $T_1$  ( $\Delta T = 30$  min, aligned with precipitation data), identify MCSs at each time step, denoted them as set  $\{A_n\}$  and  $\{B_n\}$ , respectively.

Step 2—Motion vector estimation. Use the optical flow method to calculate the average motion vectors of  $\{A_n\}$  from  $T_0$  to  $T_1$ , denoted as set  $\{\vec{v_n}\}$ . Based on these vectors, compute the theoretical positions of  $\{A_n\}$  at  $T_1$ , forming a predicted MCS set  $\{A'_n\}$ .

Step 3—Spatiotemporal mapping. For a specific MCS  $A_1 \in \{A_n\}$ , identify all elements from  $\{B_n\}$  overlapping with  $A_1'$  to form a subset  $\{B_n'\}$ . Subsequently, locate elements in  $\{A_n'\}$  that overlap with  $\{B_n'\}$ , establishing the spatiotemporal mapping relationships of  $A_1$  from  $T_0$  to  $T_1$ .

Step 4—Evolution type classification. Based on the mapping relationships, classify  $A_1$  into evolution categories according to criteria outlined in Table 1. Four primary types are defined: continuous, merging, splitting, and complex, along with four secondary subtypes. Specifically, merging, splitting, and complex types are collectively termed discontinuous MCSs.

### MCS-related precipitation

The precipitation data used in this study are derived from the GPM (Global Precipitation Measurement) gridded precipitation product, IMERG (Integrated Multi-satellitE Retrievals for GPM). IMERG includes three versions: early run, late run, and final run, with a temporal resolution of 0.5 h and a spatial resolution of 0.05°  $\times$  0.05° ( $\sim$ 5 km). The final run product is selected for this analysis  $^{30}$ . Each data from IMERG final run has a unique start time  $T_0$  and end time  $T_1$ , and the given precipitation variable reflects the average rain rate between  $T_0$  and  $T_1$ . Therefore, IMERG data are matched to Himawari-8 AHI data near the start time  $T_0$  to show the average rain rate during the process of cloud evolution.

For a given precipitation grid, it is classified as MCS-related precipitation if it occurs within 50 km of an MCS. If multiple MCSs fall within 50 km of the grid (a rare occurrence), the precipitation is assigned to the nearest MCS.

In this study, the four primary MCS evolution types account for 62.14% of the total MCS population but contribute 97.15% of the total MCS-related precipitation. In contrast, the four secondary MCS types (Sub-continuous, Sub-merging, Sub-splitting, and Dissipation) constitute 37.86% of the total MCS population yet contribute only 2.85% of the total precipitation. This disparity arises for two reasons. Firstly, secondary-type MCSs are typically small in scale and near dissipation, resulting in fewer occurrences in each location (see Fig. S1 in the Supplementary File). Secondly, secondary-type MCSs exhibit declining cloud-top heights and weak convective activity, leading to minimal rainfall production (see Fig. S2 in the Supplementary File).

Besides, to strengthen the analytical framework, this study additionally incorporates two supplementary datasets, including the ERA5 monthly reanalysis data with  $0.25^{\circ} \times 0.25^{\circ}$  and Terra MODIS Level-3 monthly AOD products with  $1^{\circ} \times 1^{\circ}$  resolution.

Table 1 | Classification criteria for the evolutionary type of MCS  $A_1$  at time  $T_0$ 

Primary types	Continuous	Merging	Splitting	Complex
Mapping	1–1	<i>n</i> −1 ( <i>n</i> > 1)	1–n (n > 1)	n−n (n > 1)
Concept <sup>a</sup>	$A_1'$	$A_1'$ $B_1$ $A_2'$	$A_1'$ $B_2$	$A_1'$ $B_2$ $A_2'$
Overlap <sup>b</sup>	>50%	>50%	>50%	0-100%
Proportion <sup>c</sup>	33.89%	14.47%	4.91%	8.87%
Rainfall contribution	28.24%	21.15%	14.53%	33.13%
Secondary types	Sub-continuous	Sub-merging	Sub-splitting	Dissipation
Secondary types  Mapping	Sub-continuous 1–1	Sub-merging n-1 (n > 1)	Sub-splitting 1-n (n > 1)	Dissipation 1–0
				-
Mapping	1-1 B <sub>1</sub>	n-1 (n > 1)  A' <sub>1</sub> B <sub>1</sub>	$1-n (n > 1)$ $B_1$ $A'_1$	1-0 A' <sub>1</sub>
Mapping Concept	1-1 B <sub>1</sub>	$n-1 (n > 1)$ $A'_1$ $A'_2$	$ \begin{array}{c} 1-n (n > 1) \\ B_1 \\ A'_1 \\ B_2 \end{array} $	1-0  A' <sub>1</sub> B <sub>1</sub>

Blue area indicates the theoretical positions of the MCS at time  $T_1$  (predicted from  $T_0$ ), while green represents the actual MCS positions observed at  $T_1$ .

## Data availability

The IMERG precipitation data was collected from the Precipitation Measurement Mission website (https://pmm.nasa.gov). The Himawari-8 AHI L1 gridded data was provided by the Japanese Meteorological Agency (https://www.data.jma.go.jp/mscweb/en/himawari89/space\_segment/spsg\_ahi.html). The ERA5 reanalysis data was collected from the ECMWF website (https://www.ecmwf.int/).

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<sup>&</sup>lt;sup>b</sup>Overlap (ratio) refers to the proportion of the overlapping area between  $A'_1$  and  $\{B_n\}$ , relative to the area of  $A'_1$  itself.

<sup>&</sup>lt;sup>c</sup>Proportion denotes the percentage of MCSs of this type relative to the total MCS population.

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### **Author contributions**

A.Z. and Y.C. discussed the central concepts and ideas and processed and analyzed data. A.Z. wrote this paper.

# **Competing interests**

The authors declare no competing interests.

### Additional information

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