



1 **Understanding the variation of Reflected Solar Radiation:** 2 **A Latitude- and month-based Perspective**

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7 **Abstract.** The hemispheric symmetry of planetary albedo (PA) is crucial for the Earth's energy budget.
8 However, our understanding of hemispheric albedo is still limited, particularly regarding its variations at
9 finer spatial and temporal scales. Using 21 years of radiation data from CERES-EBAF, this study
10 quantifies the contribution rates of different latitudes to the hemispheric reflected solar radiation and
11 examines their seasonal variations. Statistical results show that the northern latitudinal zones of 0° to 40°
12 contribute more reflected radiation than the corresponding southern latitudes, but the southern latitudinal
13 zones of 50° to 90° compensate for this. From the equator to 40°, the latitudinal contribution to the
14 hemisphere is high in autumn and winter and low in spring and summer; however, after 50°, the situation
15 is reversed. And even during extreme cases, anomalies of the cloud component contribution play a
16 dominant role in anomalies of the total reflected radiation contribution of the latitudinal zone in most
17 latitudinal zones. Additionally, this study evaluates the performance of four radiation data (including:
18 satellite and reanalysis data) in reproducing hemisphere albedo and its hemispheric symmetry compared
19 to CERES-EBAF data. Under different symmetry criteria, the applicability of different datasets to
20 hemispheric symmetry of PA studies varies. Note that the Cloud_cci AVHRR performs better in
21 capturing hemispheric symmetry. However, none of these datasets can decompose the different
22 components of reflected radiation well. These results contribute to advancing our understanding of
23 hemispheric symmetry variations and compensation mechanisms, reducing the uncertainty of model
24 simulations, and improving algorithms for different radiation datasets.

25 **1 Introduction**

26 Planetary albedo (PA) refers to the fraction of incoming solar radiation that is reflected back into
27 space by the Earth's atmosphere, clouds, and surface. It plays a crucial role in regulating the Earth's



28 energy budget and global climate change (Wielicki et al., 2005; Stephens et al., 2015) by determining the
29 amount of solar energy absorbed and distributed throughout the Earth-atmosphere system (Fu et al., 2000;
30 Stephens et al., 2015). Studies have shown that a 5% change in PA can lead to an average global
31 temperature change of approximately 1K (North et al., 1981), while a 0.01 change in PA can have a
32 radiative forcing effect equivalent to doubling the amount of carbon dioxide in the atmosphere (Wielicki
33 et al., 2005; Bender et al., 2006). Even small variations in PA can have significant implications for the
34 development of Quaternary glaciations (Budyko, 1969). Therefore, it is crucial to quantify the basic
35 statistical properties of PA, as well as understand the major principles and mechanisms governing its
36 spatial-temporal changes and long-term trends at various scales, including annual, global, and even finer
37 spatial-temporal scales (e.g., regional and monthly scales).

38 Nowadays, satellite data and model simulations have been widely utilized to investigate the
39 climatology (George and Bjorn, 2021; Jönsson and Bender, 2022), spatial and temporal distribution
40 characteristics (Loeb et al., 2007; Pang et al., 2022), and long-term trends of PA (Diamond et al., 2022;
41 Stephens et al., 2022; Xiao et al., 2023) , as well as the contributions of different components (such as
42 cloud, clear-sky atmosphere, and surface) to PA (Stephens et al., 2015; Jönsson and Bender, 2022). Long-
43 term satellite records have indicated that the current PA maintains a relatively stable value of
44 approximately 0.29 (Bender et al., 2006). Surprisingly, the annual mean reflected solar radiation in the
45 Northern Hemisphere (NH) and Southern Hemisphere (SH) is nearly the same within measurement
46 uncertainty, which is referred to as hemispheric symmetry (Loeb et al., 2009; Voigt et al., 2013; Stephens
47 et al., 2015; Jönsson and Bender, 2022). However, although satellite observations have demonstrated the
48 symmetry of hemispheric PA on inter-annual scales, state-of-the-art models still struggle to reproduce
49 this essential feature due to inadequate representation of the underlying physical mechanisms for PA
50 variation, particularly the poor modeling of compensatory effects of asymmetric clouds (Voigt et al.,
51 2013; Stephens et al., 2015; Jönsson and Bender, 2022). As a result, mean hemispheric asymmetries
52 persist in all-sky reflections from CMIP phase 3 to CMIP phase 6, with considerable spread among the
53 General Circulation Models (GCMs) within each CMIP phase (Crueger et al., 2023). Additionally,
54 models also fail to capture the observed decreasing trend in reflected shortwave radiation in both
55 hemispheres. These limitations may stem from the inability of models to accurately simulate the
56 components of PA and their respective contributions to the hemispheric symmetry of PA. In particular,



57 the annual mean reflected solar radiation at the hemispheric scale is comprised of the reflected radiation
58 at finer spatial and temporal scales (such as regional and monthly scales). If models cannot accurately
59 account for the contribution of each component to hemispheric albedo at finer temporal and spatial scales,
60 it hinders our ability to identify potential regional maintenance or compensation mechanisms for
61 hemispheric symmetry in PA. Furthermore, it introduces significant uncertainties in model simulations
62 of PA at annual and hemispheric scales.

63 Indeed, the contributions of different latitudinal zones to PA vary significantly spatially and
64 temporally due to variations in water vapor, aerosols, vegetation, and clouds (Hu and Stamnes, 1993;
65 Loeb et al., 2007; Voigt et al., 2014; Letu et al., 2018; Li et al., 2018; Zhao et al., 2019; Yang et al., 2020).
66 These changes in contribution are also influenced by local climate and large-scale circulation patterns.
67 For example, as one of the important conjectures of the compensating mechanism for the hemispheric
68 symmetry of PA (Voigt et al., 2013; Voigt et al., 2014; Stephens et al., 2015), the Intertropical
69 Convergence Zone (ITCZ) plays a role in regulating cloudiness in the 10°S-10°N region, with its location
70 and intensity varying seasonally (Hu et al., 2007). However, the presence of tropical clouds alone may
71 not be the primary factor determining hemispheric albedo symmetry, as the maximum in tropical
72 cloudiness is located in the NH along with higher contributions from the surface and clear-sky
73 atmosphere of the NH (Jönsson and Bender, 2023). Instead, extra-tropical cloudiness, particularly in the
74 SH, has been highlighted as an important factor in maintaining the symmetry of the annual mean
75 hemispheric albedo (George and Bjorn, 2021; Rugenstein and Hakuba, 2023). Recent studies have
76 emphasized the impact of the distinct land-sea distribution between hemispheres, which leads to
77 enhanced oblique pressure activity at mid-latitudes in the SH (Hadas et al., 2023). This activity results
78 in intensified storm tracks, increased cloud cover, and higher cloud albedo in the extratropical regions of
79 the SH (George and Bjorn, 2021). These clouds effectively compensate for the asymmetries in clear-sky
80 albedo between the NH and SH. The oblique pressure activity at mid-latitudes exhibits a distinct seasonal
81 cycle, with winter storm tracks in the NH being almost three times longer than summer storm tracks, and
82 seasonal meridional shifts occurring in the SH (Verlinden et al., 2011). It is important to note that the
83 contributions of different latitudinal zones to hemispheric PA are not independent of each other. Changes
84 in the contributions of different latitudinal zones can offset or amplify each other, resulting in an energy
85 balance or imbalance between the two hemispheres. For example, anthropogenic emissions from Asia



86 not only enhance the local clear-sky atmospheric component of reflected radiation through direct aerosol
87 effects but also significantly increase aerosol optical thickness in the northwestern Pacific through long-
88 range transport. This, in turn, increases the amount of deep convective clouds due to the indirect effects
89 of aerosols. The increased deep convective clouds can strengthen the storm track in the Pacific Ocean
90 and increase the contribution of the cloud component (Wang et al., 2014). Therefore, a comprehensive
91 analysis of the contributions of different components at different latitudes and their monthly variations
92 would help to better understand the mechanism of hemispheric PA symmetry and reduce uncertainties in
93 model simulations of PA.

94 Currently, satellite remote sensing products from the CERES mission, which are based on broad-
95 band measurements, are invaluable for studying the energy balance of the Earth-atmosphere system
96 (including changes in PA and hemispheric symmetry) and climate change (Loeb et al., 2018b). However,
97 the relatively limited observation record of CERES (2000-present) makes it challenging to study
98 hemispheric PA symmetry on longer time scales. In recent years, satellite radiometric products and
99 reanalysis data with longer time coverage and finer spatial resolution have been released, and numerous
100 assessments have been conducted by researchers (Cao et al., 2016; Schmeisser et al., 2018; Loeb et al.,
101 2022). The Cloud_cci version 3 radiative flux dataset has been shown to be in good agreement with the
102 CERES EBAF dataset at a global scale (Stengel et al., 2020). Zhao et al. (2022) systematically assessed
103 the applicability and accuracy of the Cloud_cci radiative flux dataset over the Tibetan Plateau (TP) and
104 found that although the AVHRR can better describe the spatial and temporal characteristics of top-of-
105 atmosphere (TOA) radiative fluxes over the TP, it does not capture the long-term trend of cloud radiative
106 effects well. Furthermore, the spatial and temporal distributions of global TOA reflected radiance from
107 MERRA-2 and ERA5 have been compared with those from CERES (Lim et al., 2021), revealing that
108 ERA5 shows better agreement with CERES than MERRA-2 in terms of seasonal fluxes. However, most
109 of these assessments focus on the spatial and temporal reproducibility of these data in terms of global or
110 regional radiative flux, while their performance in terms of hemispheric symmetry remains unknown. To
111 understand the mechanisms maintaining hemispheric symmetry of PA on longer time scales, it is essential
112 to systematically evaluate the usability of long-term radiative flux products in describing hemispheric
113 differences in TOA reflected radiation and its components.

114 To enhance future investigations into the potential maintenance mechanisms of hemispheric



115 symmetry and to reduce uncertainties in model simulations, this study aims to utilize long-term satellite
116 observations of radiative flux (e.g., CERES-EBAF ed4.2) to quantify the contributions of clear-sky
117 atmospheric, surface, and cloud components to PA at finer spatial-temporal scales (e.g., regional and
118 monthly scales). Additionally, we aim to analyze the spatial-temporal variability characteristics of these
119 contributions. Furthermore, we will examine the variation of controlling factors in extreme anomaly
120 years of reflective radiation contribution at different latitudes. In particular, we will comprehensively
121 evaluate the reproducibility of inter-hemispheric reflected radiation and its components using various
122 satellite and reanalyzed radiation datasets (including Cloud_cci AVHRR PM v3, ISCCP-FH, MERRA-
123 2, and ERA5), with a focus on comparing them to CERES EBAF radiation products. The paper is
124 structured as follows: Section 2 describes the data and methods used in the study; Section 3 presents the
125 overall characterization, latitudinal zone distribution, seasonal variations, and driving factors of reflected
126 radiative flux and its components in extreme anomaly years, as well as the assessment of different
127 radiation datasets; and finally, Section 4 provides the conclusions.

128 **2 Datasets and Methodology**

129 **2.1 Datasets**

130 **2.1.1 CERES and MODIS**

131 The Terra and Aqua satellites of the NASA were launched into Earth orbit in 1999 and 2002,
132 respectively. Here, we use the products from two instruments (Clouds and the Earth's Radiant Energy
133 System (CERES) and Moderate-Resolution Imaging Spectroradiometer (MODIS)) flying on both the
134 Terra and Aqua satellites to provide the monthly mean radiative flux, cloud properties and parameters of
135 the underlying surface.

136 CERES utilizes satellite-based observations to measure the Earth's radiation budget and clouds
137 (Wielicki et al., 1996; Loeb et al., 2018b). The CERES instrument is a scanning broadband radiometer
138 that captures radiation data across three channels: the shortwave channel (0.3–5 μ m), the infrared window
139 channel (8–12 μ m), and the total channel (0.3–200 μ m). In our study, we focus on filtered radiances within
140 the shortwave spectrum (0.3–5 μ m). These radiances are first converted to unfiltered radiance (Loeb et
141 al., 2001), and then transformed into TOA radiation flux using an empirical angular distribution model
142 (Su et al., 2015). Instantaneous fluxes are converted to daily-averaged fluxes using sun-angle dependent



143 diurnal albedo models (Loeb et al., 2018b). Surface irradiances are independently calculated using
144 aerosols, clouds, and thermodynamic properties derived from satellite observations and reanalysis
145 products. These calculations are constrained by the TOA irradiance (Kato et al., 2013; Kato et al., 2018).

146 Based on the work of Stephens et al. (2015) and Jönsson and Bender (2022), we have selected the
147 TOA and surface shortwave (SW) radiative fluxes from the CERES EBAF product to analyze the
148 contributions of different components. The CERES EBAF product employs an objectively constrained
149 algorithm (Loeb et al., 2009) that adjusts the TOA SW and longwave (LW) fluxes within their
150 uncertainties to eliminate inconsistencies between the global mean net TOA fluxes and the heat storage
151 in the Earth-atmosphere system (Johnson et al., 2016). We use CERES EBAF, edition 4.2 (Loeb et al.,
152 2018b), for monthly mean radiative fluxes (incoming solar radiation, upwelling SW radiation at TOA,
153 and both upwelling and downwelling SW radiation at the surface) during all-sky and clear-sky conditions
154 between March 2001 and February 2022 (21 years) on a $1^\circ \times 1^\circ$ resolution grid. Note that EBAF data
155 prior to June 2002 are Terra records only. In order to minimize flux discontinuities between the Terra-
156 only record and the Terra&Aqua record, the CERES EBAF Ed4.2 product applies regional climate
157 adjustments to the Terra-only record.

158 The cloud property parameters (Cloud Fraction (CF), Cloud Visible Optical Depth (CVOD), Ice
159 Water Path (IWP) and Liquid Water Path (LWP)) for this study from March 2001 to February 2022 are
160 obtained from CERES_SSF1deg_Ed4.1. This product provides MODIS-derived cloud properties and
161 auxiliary data from the observed transient footprint SSF Ed4A fluxes and clouds (Minnis et al., 2020).
162 Daytime cloud properties are calculated from the MODIS visible and infrared channels. To keep
163 consistency with the EBAF data and to minimize the errors due to the diurnal cycle of clouds, we have
164 averaged the cloud properties data from the Terra and Aqua sources over their overlapped time period
165 (2002.7-2022.2). However, since Aqua data are not available prior to July 2002, this may result in a little
166 bit uncertainty. In addition, the Ice/snow Coverage (ISC) data of CERES SSF is also used in this study.
167 Snow and ice daily coverage is derived from the NSIDC (National Snow and Ice Data Center) near real-
168 time SSM/I-SSMIS EASE-Grid Daily Global Ice Concentration and Snow Extent products, which are
169 then interpolated to monthly average products.

170 In addition to snow and ice coverage, local vegetation coverage is also one of the important factors
171 to influence the change of land surface albedo (Betts, 2000; Sandholt et al., 2002). To indicate vegetation



172 greenness, the Normalized Difference Vegetation Index (NDVI) used in this study is the Terra MODIS
173 NDVI 0.05-degree monthly product (MOD13C2 collection061), which is based on spatial and temporal
174 averages of 16-day 1-kilometre NDVI. The NDVI is calculated as the ratio between TOA reflectance of
175 a red band around 0.66 μ m and a near-infrared (NIR) band around 0.86 μ m. In general, higher values of
176 NDVI indicate a higher density of green vegetation.

177 **2.1.2 ISCCP-FH**

178 The International Satellite Cloud Climatology Project (ISCCP) aims to provide global cloud
179 coverage and cloud radiation characteristics (Schiffer and Rossow, 1983). As part of the ISCCP project,
180 the ISCCP-FH radiation product contains SW radiation fluxes at five levels from the surface to the TOA
181 (surface-680hPa-440hPa-100hPa-TOA) under all-sky, clear-sky and overcast-sky conditions as well as
182 the diffuse and direct SW fluxes at the surface. ISCCP-FH is not produced using direct instrumental
183 observations, but rather the ISCCP H series of data products that are derived from geostationary and
184 polar-orbiting satellites (Young et al., 2018), adopting a complete radiative transfer model developed
185 from the GISS GCM ModelE. As a third-generation product, ISCCP-H has become more advanced and
186 has other improvements in radiation quality control, calibration, cloud detection (especially high clouds,
187 thin clouds and polar clouds), cloud and surface properties retrievals (Zhang et al., 2023). The ISCCP-
188 FH product consists of five sub-products, of which the PRF (surface-to-TOA flux profile) sub-product
189 can provides 34 years of global radiative flux data from July 1983 to June 2017 with a spatial resolution
190 of up to 1° and a temporal resolution of 3 hours. In order to be consistent with CERES EBAF data, this
191 study uses the diurnal mean of monthly mean of 3-hour upward and downward SW radiative flux at the
192 TOA and surface under all-sky and clear-sky conditions provided by the MPF (monthly average of PRF)
193 sub-product.

194 **2.1.3 AVHRR**

195 The Cloud_cci project is part of the European Space Agency (ESA) Climate Change Initiative (CCI)
196 program and aims to provide a long-term and consistent cloud property dataset (Hollmann et al., 2013).
197 The Cloud_cci dataset is generated using the state-of-the-art retrieval system called "the Community
198 Cloud retrieval for Climate" (CC4CL), which employs optimal estimation (OE) techniques and is applied
199 to passive imaging sensors from current and past European and non-European satellite missions (Sus et



200 al., 2018). In our study, we have chosen the Cloud_cci AVHRR-PMv3 dataset, which offers
201 comprehensive cloud and radiation flux properties on a global scale from 1982 to 2016. These properties
202 are derived from measurements obtained by the Advanced Very High Resolution Radiometer (AVHRR)
203 instrument onboard the afternoon (PM) satellite of the US National Oceanic and Atmospheric
204 Administration's (NOAA) Polar Operational Environmental Satellite (POES) mission (Stengel et al.,
205 2020). To account for the diurnal cycle of the solar zenith angle, all samples of the SW flux are rescaled
206 and averaged to represent a 24-hour average for each pixel. This monthly average value is then
207 determined (More details can be found in ESA Cloud_cci Algorithm Theoretical Baseline Document
208 v6.2). It is important to note that the radiation broadband flux is determined by combining exported cloud
209 characteristics with reanalysis data (Stengel et al., 2020). However, there are some discrepancies in this
210 product for the years 1994 and 2000 due to the unavailability of AVHRR data. Therefore, data from these
211 years are not utilized in our study. We use the monthly mean global 0.5° grid data (Level-3C) from
212 Cloud_cci, which includes TOA and surface upward and downward SW radiative fluxes under both all-
213 sky and clear-sky conditions. We interpolate this data to a 1° grid to maintain consistency with CERES.

214 **2.1.4 Reanalysis datasets**

215 The Modern-Era Retrospective Analysis for Research and Applications, version 2 (MERRA-2), is
216 the latest atmospheric reanalysis of the modern satellite era produced by NASA's Global Modeling and
217 Assimilation Office (GMAO) with version 5.12.4 of the Goddard Earth Observing System (GEOS)
218 atmospheric data assimilation system (Gelaro et al., 2017). It is the first long-term global reanalysis to
219 assimilate space-based observations of aerosols and represent their interactions with other physical
220 processes in the climate system. MERRA-2 can provide long-term radiative and aerosol products with a
221 spatial resolution of 0.5°×0.625° from 1980. Here, the radiative product from MERRA-2 is used for
222 comparative assessment with CERES data, as well as the aerosol product and land surface product are
223 used as the main influence factors of the clear-sky atmospheric component and surface component of the
224 PA, respectively. M2TMNXRAD (or tavgM_2d_rad_Nx) monthly mean radiation flux data, including
225 the incident and net downward SW radiative fluxes at the TOA and the surface under all-sky and clear-
226 sky conditions, M2TMNXAER (or tavgM_2d_aer_Nx) 550nm total aerosol extinction optical depth
227 (AOD) monthly data, and M2TMNXLND (or tavgM_2d_Ind_Nx) surface soil moisture (SM) monthly
228 data from 2001 to 2021 are selected in this investigation.



229 ERA5 is the fifth-generation atmospheric reanalysis of the global climate from January 1940 to
230 present by the European Centre for Medium-Range Weather Forecasts (ECMWF). ERA5 combines
231 model data with observations from around the world to form a globally consistent dataset that replaces
232 the previous ERA-Interim reanalysis. 4D-var data assimilation technique in the Integrated Forecasting
233 System (IFS) Cycle 41r2 is used to ensure a significant improvement in prediction accuracy and
234 computational efficiency (Jiang et al., 2019; Hersbach et al., 2020). It provides hourly estimates of a
235 large number of atmospheric, land and oceanic climate variables with a spatial resolution of $0.25^\circ \times 0.25^\circ$
236 (Hersbach et al., 2020). The monthly average surface and TOA radiation budget products are used in this
237 study. The total column water vapor (TCWV) data also come from ERA5.

238 In order to ensure data consistency, we resample the monthly mean diurnal averaged radiative fluxes
239 and other meteorological parameters from MERRA-2 and ERA5 datasets to match the $1^\circ \times 1^\circ$ resolution
240 of CERES.

241 It's important to note that for a more accurate comparison with CERES EBAF, we utilize the other
242 radiative flux data mentioned above (SW radiative flux from ISCCP-FH, AVHRR, ERA5, and MERRA-
243 2) for the overlapping time period of March 2001 to February 2016.

244 2.2 Methodology

245 2.2.1 Decomposition of planetary albedo contribution

246 To investigate the main drivers of the PA, we use the similar model as Stephens et al. (2015) to
247 decompose the PA into the contributions of the surface and atmospheric components. Assuming that
248 surface and atmospheric scattering have isotropy, planetary albedo R is defined as:

$$249 R = \frac{F_{\text{TOA}}^\uparrow}{S} \quad (1)$$

250 Among them, the F_{TOA}^\uparrow is reflected SW (upwelling) flux at the TOA, S is the solar incoming
251 (downwelling) flux. The transmittance T of the whole earth-atmosphere system is defined as:

$$252 T = \frac{F_S^\downarrow}{S} \quad (2)$$

253 Where, F_S^\downarrow is the downwelling SW radiation at the surface. The surface albedo α is calculated as
254 follows:

$$255 \alpha = \frac{F_S^\uparrow}{F_S^\downarrow} \quad (3)$$

256 Where F_S^\uparrow is the upwelling SW radiation at the surface. The term F_S^\downarrow can be expressed as:



257
$$F_S^\downarrow = tS + rF_S^\uparrow \quad (4)$$

258 Here, r and t represent atmospheric intrinsic reflectivity (that is, PA purely contributed by the
259 atmosphere when surface albedo is assumed to be 0) and atmospheric transmittance, respectively. The r
260 and t are calculated separately, so absorption and forward scattering are included in t . F_{TOA}^\uparrow can be
261 represented as:

262
$$F_{\text{TOA}}^\uparrow = rS + tF_S^\uparrow \quad (5)$$

263 By combining the above equations, R and T can be expressed by r , t and α :

264
$$R = r + \frac{\alpha t^2}{1 - r\alpha} \quad (6)$$

265
$$T = \frac{t}{1 - r\alpha} \quad (7)$$

266 According to the above equation, the values of r and t can be written:

267
$$r = R - t\alpha T \quad (8)$$

268
$$t = T \frac{1 - \alpha R}{1 - \alpha^2 T^2} \quad (9)$$

269 It can be seen that the planetary albedo R is composed of two parts: atmospheric contribution r and
270 surface contribution $\frac{\alpha t^2}{1 - r\alpha}$. These two parts are multiplied by the incoming solar radiation flux S
271 respectively, and the respective contribution values of the atmosphere and the surface to the SW
272 upwelling flux at the TOA (F_{TOA}^\uparrow) can be obtained, namely F_{atm}^\uparrow and F_{surf}^\uparrow (unit: W m^{-2}).

273
$$F_{\text{atm}}^\uparrow \equiv Sr \quad (10)$$

274
$$F_{\text{surf}}^\uparrow \equiv S \frac{\alpha t^2}{1 - r\alpha} \quad (11)$$

275 Following Jönsson and Bender (2022), we further decompose the atmospheric component into
276 clear-sky atmospheric and clouds contributions. The difference between the all-sky atmospheric
277 contribution F_{atm}^\uparrow and the clear-sky atmospheric contribution $F_{\text{atm,clear}}^\uparrow$ is considered as the cloud
278 contribution $F_{\text{cloud}}^\uparrow$. That is,

279
$$F_{\text{TOA}}^\uparrow = F_{\text{atm}}^\uparrow + F_{\text{surf}}^\uparrow = F_{\text{cloud}}^\uparrow + F_{\text{atm,clear}}^\uparrow + F_{\text{surf}}^\uparrow \quad (12)$$

280
$$F_{\text{cloud}}^\uparrow = F_{\text{atm}}^\uparrow - F_{\text{atm,clear}}^\uparrow \quad (13)$$

281 2.2.2 Regional mean and contribution rate

282 The regional averaged TOA reflected SW flux F_k is spatially aggregated using the following
283 calculation formula (Huang et al., 2012):



284
$$F_k = \frac{\sum_{i=1}^{N_k} W_{ki} \cdot F_{ki}}{\sum_{i=1}^{N_k} W_{ki}} \quad (14)$$

285 Here, N_k is the number of grid samples in region k, and F_{ki} is the reflected SW flux

286 corresponding to grid i in the region k. Here, $W_{ki} = \cos(\frac{\theta_i \pi}{180.0})$, where θ_i is the latitude of grid i.

287 Regional averages for other variables are calculated according to the similar weighting equation.

288 In order to explore the contribution of different regions to the total reflected radiation at a
289 hemispheric scale, the global latitude is divided into 18 latitude zones in the unit of 10°, that is, 90° N-
290 80° N, 80° N-70° N, ..., 70° S-80° S, 80° S-90° S. For example, the contribution rate of the cloud
291 component C_{cloud} of each latitude zone to its hemispheric reflected solar radiation can be calculated by
292 the following formula:

293
$$C_{cloud} = \left(\frac{total_latzone_cloud}{total_hem_R} \right) \times 100\% \quad (15)$$

294 Here, $total_latzone_cloud$ and $total_hem_R$ are calculated from the molecular part of Eq. (14).

295 Where $total_latzone_cloud$ refers to the sum of the cloud contribution of reflected solar radiation
296 from all grid points in the given latitude zone, $total_hem_R$ is the sum of total reflected solar radiation
297 from all grid points in the hemisphere in which the latitude zone is located (both taking into account the
298 geodetic latitudinal weights of the different grids). The contribution of surface and clear-sky atmospheric
299 components to hemispheric reflected radiation in different latitudinal zones can be derived by the similar
300 method.

301 2.2.3 Time average

302 For the average contribution over time, we consider March to the following February as a complete
303 year. Following the CERES EBAF Ed4.1 Data Quality Summary (2020), the monthly average data is
304 weighted by the number of days in each month to obtain the annual average data (Wielicki et al., 1996;
305 Loeb et al., 2009; Rugenstein and Hakuba, 2023). For example, the annual average value of TOA SW
306 reflected radiation flux in a certain year is:

307
$$F_{year} = \sum_{i=1}^{i=12} \frac{DAY_{mon}(i)}{DAY_{year}} F_{mon}(i) \quad (16)$$

308 where DAY_{year} is the total number of days in the given year, $DAY_{mon}(i)$ is the number of days
309 in the current month, and $F_{mon}(i)$ is the monthly averaged radiation flux. The annual average values of
310 all variables are also obtained by this method.



311 **2.2.4 CCHZ-DISO data evaluation system**

312 To find out whether other radiation datasets can exhibit the similar hemispheric symmetry of PA,
313 the CCHZ-DISO data evaluation system is also used. This method uses the Euclidean Distance between
314 indices of simulation and observation (DISO) to evaluate the combined quality or overall performance
315 of data from different models (Hu et al., 2019; Zhou et al., 2021; Hu et al., 2022). DISO has the advantage
316 of quantifying the combined accuracy of different models compared to Taylor diagram (Kalmár et al.,
317 2021). Moreover, the statistical indicators chosen for the Taylor diagram are fixed, whereas those in
318 DISO can be taken and discarded according to the needs of the study (Hu et al., 2022). In particular,
319 Taylor diagrams only provide statistical metrics on two-dimensional plots, DISO not only provides
320 distances in three-dimensional space to quantify the comprehensive performance of a simulation model,
321 but also allows a single statistical metric to capture different aspects of model performance (Hu et al.,
322 2019).

323 In this paper, the difference of TOA reflected radiation flux between the NH and SH from CERES
324 during 2001.3-2016.2 is taken as the observed data, while AVHRR, ISCCP, MERRE-2, ERA5 are
325 considered as the model dataset. The correlation coefficient (CC), normalized absolute error (NAE) and
326 normalized root mean square error (NRMSE) are used to construct the CCHZ-DISO 3D evaluation
327 system. The smaller the value of DISO, the closer this model dataset is to the observed data, i.e., the
328 better its composite performance.

329
$$CC = \frac{\sum_{k=0}^n (a_i - \bar{a})(b_i - \bar{b})}{\sqrt{\sum_{k=0}^n (a_i - \bar{a})^2} \sqrt{\sum_{k=0}^n (b_i - \bar{b})^2}} \quad (17)$$

330
$$AE = \frac{1}{n} \sum_{k=0}^n (b_i - a_i) \quad (18)$$

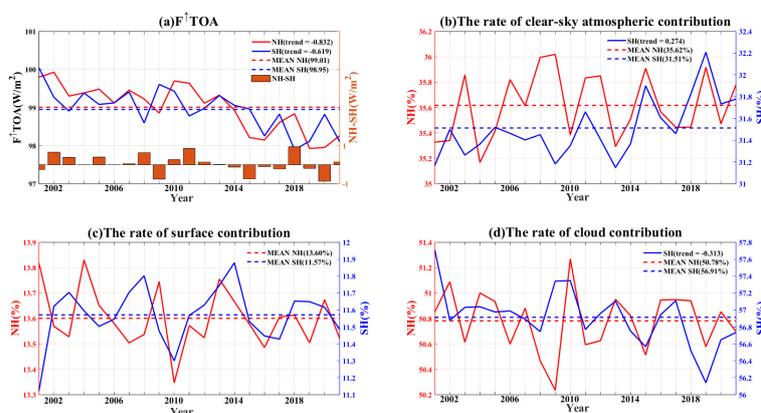
331
$$RMSE = \sqrt{\frac{1}{n} \sum_{k=0}^n (b_i - a_i)^2} \quad (19)$$

332
$$DISO_i = \sqrt{(CC_i - CC_0)^2 + (NAE_i - NAE_0)^2 + (NRMSE_i - NRMSE_0)^2} \quad (20)$$



333 **3 Results and discussion**

334 **3.1 Temporal variation of contribution of PA components in different latitudinal zones**



335

336 **Figure 1: (a) The interannual variability of upwelling SW radiative flux at the TOA in the NH and SH from**
 337 **2001 to 2021 (the left axis) and the annual mean difference between hemispheres (orange columns on the right**
 338 **axis). Interannual variation in the contribution rate of (b) the clear-sky atmospheric component, (c) the**
 339 **surface component, and (d) the cloud component to the SW upward radiative flux at the TOA in the NH (the**
 340 **left axis) and SH (the right axis); note that the scales of the two axes are not the same. The red line is for the**
 341 **NH, the blue line is for the SH, and the dashed line is the 21-year average. The trends labeled in the upper**
 342 **right corner passes the 99% significance test in units of $W m^{-2} (10a)^{-1}$ for (a) and $\% (10a)^{-1}$ for (b)-(d).**
 343 **Unlabeled trends do not pass the test of significance.**

344 Firstly, we examine the general characteristics of reflected radiation in the NH and SH on an annual
 345 average scale. Figure 1 illustrates the interannual variability of SW upwelling radiative flux at the TOA
 346 and the contribution rates of its three components in the NH and SH during the period of 2001-2021,
 347 based on CERES EBAF data. Supplementary materials (Fig. S1) provide further details on the
 348 interannual changes of reflected radiative flux by the three components. In a recent study, George and
 349 Bjorn (2021) argued that the symmetry of albedo cannot be established on an annual or sub-annual scale,
 350 but rather on larger spatial and temporal scales. In line with previous research (Stephens et al., 2015;
 351 Jönsson and Bender, 2022), our investigation demonstrates a clear symmetry in the total reflected
 352 radiation. This symmetry is evident in both the multi-year average of reflected radiation and the long-
 353 term trend of the annual average. The difference in the annual average total reflected radiant flux between



354 the hemispheres is less than 1 W m^{-2} , and the 21-year average difference approaches zero, indicating a
355 nearly equal distribution of reflected SW radiation between the NH and SH. Figure S1d illustrates the
356 cumulative year-to-year averaging of hemispheric differences in reflected radiation. It shows that the
357 hemispheric differences in total reflected radiation and its components are decreasing or tending to
358 stabilize over time, except for the clear-sky atmospheric component. The clear-sky atmospheric
359 component exhibits a strong perturbation over time, which is closely tied to human activities, particularly
360 the highly variable emissions of anthropogenic aerosols.

361 The reflected SW radiation at the TOA in both hemispheres exhibits a consistent decreasing trend
362 ($\text{Trend_NH} = -0.832 \text{ W m}^{-2} (10\text{a})^{-1}$; $\text{Trend_SH} = -0.619 \text{ W m}^{-2} (10\text{a})^{-1}$), indicating simultaneous darkening
363 of both hemispheres as observed from space. Moreover, the interannual variability of hemispheric
364 differences is increasing, and the perturbations are intensifying. To investigate whether these trends in
365 reflected radiation are linked to changes in incident radiation, we also present the interannual variations
366 of incident solar radiation and PA (Fig. S2). The results indicate that the interannual variations of incident
367 solar radiation at TOA in both hemispheres do not exhibit a clear trend, with the hemispheric difference
368 following a stable multi-year cycle. However, PA in both hemispheres shows a consistent decreasing
369 trend ($\text{Trend_NH} = -2.4 \times 10^{-3} (10\text{a})^{-1}$; $\text{Trend_SH} = -1.8 \times 10^{-3} (10\text{a})^{-1}$), suggesting a decrease in reflected
370 solar radiation by the Earth as a whole and an increase in absorbed solar radiation. However, the same
371 response in both hemispheres is driven by different component changes. The darkening of the SH can be
372 primarily attributed to a decrease in reflected radiation from the cloud component ($-0.661 \text{ W m}^{-2} (10\text{a})^{-1}$)
373 (Fig. S1c). In contrast, the reflected radiative fluxes by all three components in the NH show a decreasing
374 trend, with the cloud component exhibiting the largest decrease ($-0.448 \text{ W m}^{-2} (10\text{a})^{-1}$), followed by the
375 clear-sky atmospheric component ($-0.219 \text{ W m}^{-2} (10\text{a})^{-1}$), and the surface component showing the
376 smallest decrease ($-0.159 \text{ W m}^{-2} (10\text{a})^{-1}$). However, there is no clear trend in the proportion of their
377 contributions over the NH (Fig. 1). The decreasing trend of reflected radiation from the surface
378 component in the NH mainly originates from changes in snowpack and sea ice at the poles, but its impact
379 on the global average appears to be insignificant (Stephens et al., 2022). Additionally, the decreasing
380 trend of reflected radiation in the NH can be attributed to reduced scattering of aerosol particles (Loeb et
381 al., 2021a; Stephens et al., 2022) and a decrease in low cloud cover (Loeb et al., 2018a; Loeb et al., 2020).
382 The decrease in low cloudiness may be linked to a shift in the Pacific Decadal Oscillation (PDO) phase

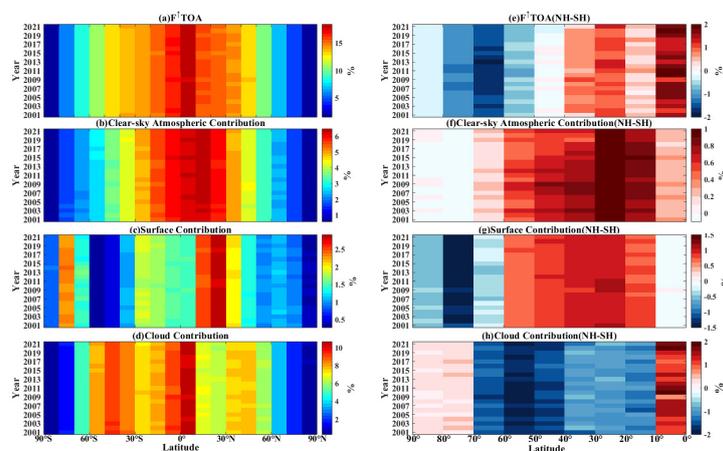


383 from negative to positive, resulting in warmer sea surface temperatures (SSTs) in parts of the eastern
384 Pacific, which significantly reduces low cloud cover and reflected solar radiation (Andersen et al., 2022).
385 To further investigate the inter-hemispheric differences in Earth's energy balance, we also calculate the
386 trends of outgoing longwave radiation and net radiation at the TOA (figure not shown). A significant
387 increasing trend of longwave radiation emitted to space is found in the NH ($0.324 \text{ W m}^{-2} (10\text{a})^{-1}$), while
388 no significant trend is observed in the SH. Loeb et al. (2021b) noted that the increase in outgoing
389 longwave radiation is primarily due to the increasing global surface temperature and changes in clouds,
390 although it is partly compensated by the increase in water vapor and trace gases. However, the overall
391 increase in outgoing longwave radiation does not outweigh the decrease in reflected shortwave radiation,
392 resulting in a positive trend in the net radiative flux in both hemispheres (indicating that the Earth is
393 absorbing more energy) (Raghuraman et al., 2021). This positive trend in the Earth's energy imbalance
394 (EEI) will exacerbate global warming, sea-level rise, increased internal heating of the oceans, and melting
395 of snow and sea ice (IPCC, 2013; Von Schuckmann et al., 2016; Loeb et al., 2021b). Indeed, a recent
396 study based on long-term homogenized radiosonde data indicated that the atmosphere has become
397 increasingly unstable in the NH during the period 1979-2020 (Chen and Dai, 2023).

398 In Fig. 1, the contributions of the three components to the PA exhibit varying degrees of hemispheric
399 asymmetry. Among the three components, clouds contribute the most, accounting for over 50%, followed
400 by the clear-sky atmosphere, and the surface with the least contribution. The cloud contribution in the
401 SH is approximately 6.13% higher than that in the NH, which can be attributed to the presence of more
402 abundant and brighter cloudiness in the SH. The clear-sky atmospheric contribution is 4.11% higher in
403 the NH compared to the SH, possibly due to greater anthropogenic aerosol emissions resulting from
404 human activities in the NH. The clear-sky atmospheric contribution rate in the SH shows an increasing
405 trend of 0.274% per decade, which may be attributed to a decrease in the cloud component contribution
406 of -0.313% per decade. This is because there is no clear trend in the reflected radiative flux by the clear-
407 sky atmosphere in the SH, but there is a significant decrease in the cloud component (Fig. S1c). However,
408 from a radiative flux perspective, the hemispheric asymmetry of the clear-sky atmosphere is decreasing
409 (Fig. S1a), primarily influenced by the declining reflection of the clear-sky atmosphere in the NH, which
410 is associated with the recent reduction of anthropogenic aerosols in eastern North America, Europe, and
411 East Asia (Raghuraman et al., 2021; Quaas et al., 2022; Stephens et al., 2022). In comparison to the SH,



412 the NH exhibits a 2.03% higher surface contribution (Fig. 1c). Although the NH has a greater land
 413 distribution, the higher ice albedo in Antarctica partially compensates for the lack of land area in the SH,
 414 resulting in a less significant difference in surface contribution between the hemispheres.

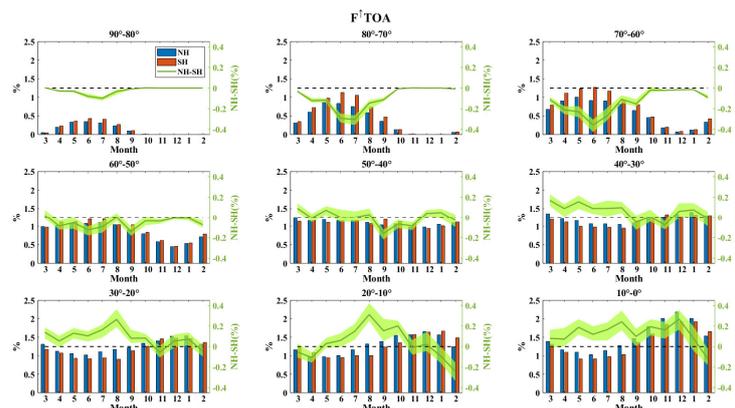


415
 416 **Figure 2: (a) Contribution of different latitudinal zones to hemispheric total reflected radiation at TOA from**
 417 **2001 to 2021 and the corresponding components: (b) the clear-sky atmospheric component, (c) surface**
 418 **component and (d) cloud component. Contribution differences between the corresponding latitudinal zones**
 419 **of the two hemispheres (NH minus SH) are also given in (e), (f), (g) and (h).**

420 Large-scale systems or certain compensatory mechanisms that may affect the hemispheric
 421 symmetry of PA do not directly operate on a hemispheric scale. Instead, they can compensate for
 422 hemispheric energy imbalances by influencing local or regional climates. For instance, oblique pressure
 423 activity, although primarily occurring at mid-latitudes, exerts a significant influence on cloud albedo,
 424 thereby strongly impacting global albedo (Hadas et al., 2023). While larger regional anomalies in
 425 reflected radiative flux may offset each other when spatially and temporally averaged to calculate global
 426 PA and its interannual variations, these anomalies play a crucial role in regional radiation budgets,
 427 subsequent climate change, and the identification of mechanisms that maintain or compensate for PA.
 428 Therefore, to gain further insight into the regional effects of these influencing mechanisms, we have
 429 divided the globe into 18 latitudinal zones in 10° increments. Figure 2 illustrates the contribution of each
 430 latitudinal zone to the total reflected radiation of its respective hemisphere, where (b)-(d) of Fig. 2 depict
 431 the contribution of the three components to the total reflected radiative energy of the hemisphere in each
 432 latitudinal zone.



433 In general, the interannual variation in the contribution rate of each latitude zone to the total
434 hemispheric reflected radiation is small. However, there are still some relatively anomalous years, which
435 will be discussed in detail in the next section (3.2). More energy is reflected from the 0°-40° latitude
436 zones in the NH compared to the corresponding latitude zones in the SH (Fig. 2c). However, this
437 imbalance is compensated by more reflection from the SH in the 50°-90° latitude zones. The dominance
438 of the NH mainly arises from the clear-sky atmospheric contribution between 0°-70°, the surface
439 contribution from 10°-60°, and the cloud contribution from the equator to 10°. In contrast, the strength
440 of the SH at middle and high latitudes is derived from the surface contribution from 60°-90° and the
441 cloud contribution from 10°-70°. Regarding clear-sky atmospheric contributions, the NH as a whole
442 slightly exceeds the SH (except in the polar regions), possibly due to the higher presence of dust aerosols
443 in the NH tropics and subtropics, as well as more sulfate pollution in the mid-latitudes (Diamond et al.,
444 2022). Notably, the disparity in clear-sky atmospheric contributions between the two hemispheres is
445 greatest at 20°-30° (around 1%), influenced by the combined effect of more dust and sulfate aerosols in
446 the NH. There are significant hemispheric differences in surface contributions, with the NH exhibiting
447 larger surface contributions concentrated in the 10°-60° latitude range, surpassing those in the SH. This
448 discrepancy is expected due to the larger land area in the NH. Conversely, in the region from 60° to the
449 poles, particularly from 70°-80°, the SH shows a greater surface contribution due to higher snow and ice
450 coverage in the near-polar regions, resulting in more solar radiation reflection. However, over the Arctic,
451 surface warming is occurring at a rate nearly four times faster than the global warming rate (Mika et al.,
452 2022), leading to a continued decrease in Arctic ice cover. The SH also exhibits a significant contribution
453 rate from clouds between 10° and 70°. This is not only attributed to lower contributions from the surface
454 and clear-sky atmosphere but also to the prevalence of subtropical cloudiness and higher cloud albedo at
455 mid-latitudes (Engström et al., 2017). The greater contribution of NH clouds near the equator may be
456 due to the persistent presence of the Intertropical Convergence Zone (ITCZ) north of the equator in the
457 eastern Pacific and Atlantic. This observation suggests that the SH heavily relies on extratropical clouds
458 to compensate for clear-sky hemispheric asymmetries.



459

460 **Figure 3: Monthly variation of the total reflected SW radiation contribution of different latitude zones to the**
 461 **total reflected SW radiation of their hemispheres and their hemispheric differences. The blue bars are for the**
 462 **NH, orange for the SH, corresponding to the left axis; the green line represents the inter-hemispheric**
 463 **difference, corresponding to the right axis, and the green shading indicates the difference spread in**
 464 **hemispheric difference for the corresponding month in that latitude zone during 2001-2021. The months are**
 465 **marked according to the NH, corresponding to the SH months of September, October, ..., January,**
 466 **February, ..., and August.**

467 The analysis presented above is based on the results of annual average reflected radiation. It is
 468 important to note that the symmetry of PA between hemispheres is a characteristic observed at interannual
 469 scales. However, certain natural and human activities that strongly influence albedo or compensate for
 470 hemispheric symmetry operate seasonally or even occur only in specific months of the year. These
 471 activities can have a significant impact on interannual scales, with their signals being more pronounced
 472 during particular seasons. We therefore hope to further clarify the variations of these mechanistic signals
 473 by resolving the reflected radiation and its components at finer temporal scale (e.g., monthly).

474 Figure 3 illustrates the monthly changes of contribution from different latitude zones to the total
 475 amount of reflected SW radiation and their contribution differences between two hemispheres. It is
 476 evident that the total reflected radiation from different latitude zones in both hemispheres exhibits
 477 noticeable monthly changes. From the equator to the 40° region, higher contributions are observed during
 478 autumn and winter, while lower contributions are seen during spring and summer. Additionally, in most
 479 months, the latitude zones in the NH have higher contributions compared to those in the SH. However,
 480 from the 50° to the polar region, the annual cycle of contribution rate reverses. Specifically, the
 481 contributions during spring and summer become more prominent in these regions. Notably, the latitude



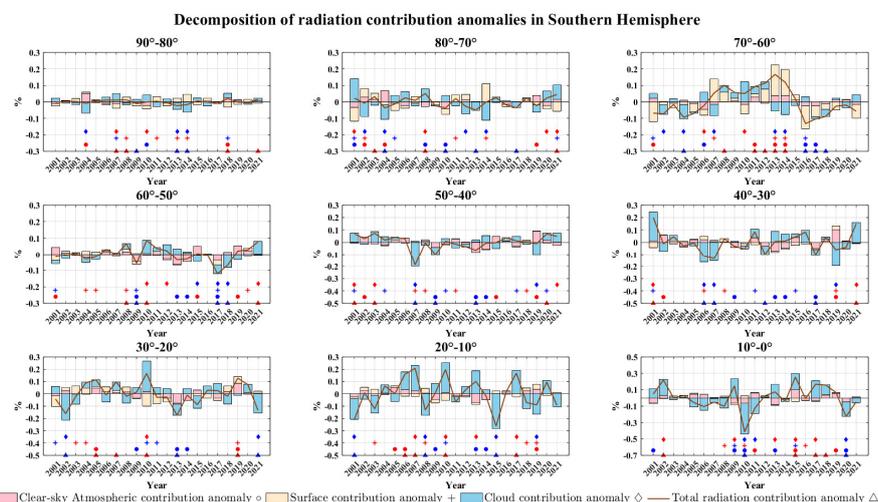
482 zones in the SH consistently exhibit higher contributions to the entire hemisphere throughout the year
483 compared to the corresponding latitude zones in the NH. This difference is particularly significant during
484 June and July in the 60°-80° latitude range, primarily due to variations in contributions from the surface
485 and cloud components (Fig. S4-S5). Overall, the annual cycle is primarily determined by seasonal
486 variations in the contributions of the cloud component (Fig. S5) and the clear-sky atmospheric component
487 (Fig. S3) at low and middle latitudes. Additionally, the surface component exerts a strong influence at
488 high latitudes (Fig. S4). The reversal of the annual cycle of the latitudinal zone contribution of reflected
489 radiation after 40° is mainly attributed to similar variability characteristics observed in incident solar
490 radiation (Fig. S6).

491 From 10° to 60°, the surface contribution in the NH consistently exceeds that in the SH, likely due
492 to the larger land area in the NH (Fig. S4). However, there is no clear monthly variation pattern for the
493 interhemispheric differences in surface contributions within these latitudinal zones. At low latitudes, the
494 surface contributions in both hemispheres exhibit similar monthly variations to those observed in the
495 total reflected radiation contribution (Fig. 3). In the 0°-10° range, which is predominantly oceanic, the
496 hemispheric difference in surface contributions is nearly negligible. From 60° to 90°, the dominant role
497 of summer in the SH becomes more pronounced, with a greater contribution from the surface component
498 compared to the cloud component at 70° S-90° S. However, in the NH, the cloud component still
499 contributes the most at 60°-90°. Furthermore, in these high-latitude zones, there is no significant annual
500 cycle in the hemispheric differences of incident solar radiation. Therefore, the hemispheric differences
501 in surface contributions are primarily influenced by surface albedo (Fig. S7). These regions are located
502 close to the poles and have extensive ice and snow cover. With global warming, ice and snow melting is
503 occurring at a rapid pace, resulting in noticeable seasonal changes in ice and snow cover. Notably, at 70°-
504 80°, the annual cycle pattern of the hemispheric difference in surface component contribution (Fig. S4)
505 closely resembles that of surface albedo (Fig. S7), albeit with the latter exhibiting extremes 2-3 months
506 later than the former. This discrepancy may arise from the fact that the contribution of the surface
507 component is defined relative to the reflected radiation at the TOA, while surface albedo is influenced
508 by cloud masking and modulation (Qu and Hall, 2005). Additionally, the surface contribution from the
509 NH at 70°-80° peaks in May, one month earlier than in the SH, although both hemispheres experience
510 their incident radiation peak in June (Fig. S6). This discrepancy is related to the distinct patterns of

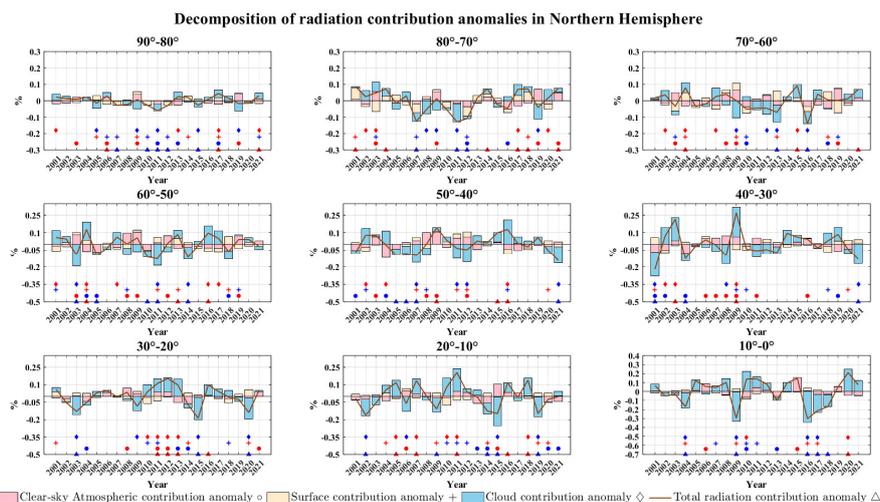


511 monthly changes in surface albedo between the two hemispheres and the differential responses of polar
 512 snow and ice to global warming. The Arctic melt season is advancing and lengthening due to global
 513 warming and the Arctic amplification effect (Noël et al., 2015; Wang et al., 2018). From a hemispheric
 514 difference perspective, the annual cycle of total reflected radiation contribution primarily stems from the
 515 cloud component contribution in mid-low latitude regions and from surface and cloud contributions in
 516 high latitude zones, with the clear-sky atmospheric contribution exhibiting relatively weak variability.
 517 The interannual spread of the annual cycle of cloud component contribution is significant across all
 518 latitudinal zones, whereas the clear-sky atmosphere and surface contributions remain relatively stable.
 519 This indicates that the interannual variability of the seasonal radiation cycle in different latitudinal zones
 520 is predominantly driven by cloud contributions.

521 **3.2 Contribution of different factors to latitudinal zones in extreme years**



522
 523 **Figure 4: Anomaly time series of contribution rate of total reflected radiation and its three components of**
 524 **each latitude zone to total hemispheric reflected radiation in the SH during 2001 to 2021. The pink is the**
 525 **contribution anomaly of the clear-sky atmosphere, the yellow is the contribution anomaly of the surface, the**
 526 **blue is the contribution anomaly of the cloud, and the brown line represents the total contribution anomaly**
 527 **to the total hemispheric reflected radiation. The triangles labeled in the figure indicate that the contribution**
 528 **anomaly of total reflected radiation in this latitudinal zone for the year exceeds one of its standard deviations,**
 529 **i.e., it is an extreme value, with blue indicating extreme lows and red indicating extreme highs. Circles, plus**
 530 **signs, and diamonds indicate extreme values of clear-sky atmospheric contribution anomalies, surface**
 531 **contribution anomalies, and cloud contribution anomalies, respectively. Note that the vertical scale is different**
 532 **for different latitudinal zones.**



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Figure 5: Same as Fig.4, but for the NH.

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Different components play distinct roles in each latitudinal zone, and this section further explores which component dominates the variation of reflected radiation contribution under extreme conditions. It would be a valuable improvement if the model could capture the anomaly in contribution from different components across latitudinal zones during such extremes. Thus, we begin by decomposing the yearly contribution anomaly of total reflected radiation for each latitude zone into contribution anomalies of the three components. Subsequently, we identify extreme values of the total reflected radiation contribution anomaly and the three component contribution anomalies for each latitudinal zone by exceeding their respective one standard deviation (indicated in Fig. 4 and Fig. 5 using different symbols). Additionally, we label years with extreme anomalous values of total reflected radiation contribution anomaly as "extreme anomalous years" (represented by triangles in Fig. 4 and Fig. 5). Considering that variations in PA and its components are influenced by atmospheric and surface properties (Loeb et al., 2007; Voigt et al., 2014; Jian et al., 2018), we select surface-related factors: ISC, SM, NDVI; cloud-related factors: LWP, IWP, CVOD, CF; and clear-sky atmospheric-related factors: TCWV, AOD. To further correlate the anomalies contributed by different components in different latitudinal zones with changes in the control factors during extreme anomalous years, we calculate the proportion of anomalies in the different factors that fall outside their normal ranges (one standard deviation) (Fig. S8, S9). This proportion is obtained by calculating the annual mean anomaly of each factor in certain latitude zone minus its standard deviation and then dividing by the standard deviation. Importantly, it should be noted that the radiation



553 contributions from different latitudinal zones exhibit varying sensitivities to changes in different factors,
554 resulting in different magnitudes of response. For instance, in the 10° S-20° S zone, snow formation is
555 challenging and limited to high altitudes in the Andes Mountains (Saavedra et al., 2017). Therefore,
556 changes in ISC in this latitudinal zone reflect changes in only a small part of the region and have little
557 effect on the contribution of reflected radiation from the entire latitudinal zone.

558 Overall, the range of interannual variation in reflected radiation contribution is relatively large in
559 the middle and low latitudes, while it remains more stable near the poles with minimal fluctuations.
560 Previous studies have demonstrated that cloud variability dominates the variability of PA (Stephens et
561 al., 2015; Seinfeld et al., 2016), and this conclusion holds true across different latitude zones. In most
562 latitude zones in both hemispheres, especially in the tropics and subtropics, the radiative contribution
563 variability is primarily influenced by the cloud component (Jönsson and Bender, 2022). This conclusion
564 also applies to extreme cases. Globally, 87 % of extreme anomalous years are dominated by contribution
565 anomalies from the cloud component, 10 % by the surface component, and 3 % by the clear-sky
566 atmospheric component (Fig. S10). However, there is a slight difference between the NH and SH. In the
567 SH, 18 % of extreme years are dominated by anomalies in the surface component contribution, compared
568 to only 3 % in the NH. Among all events with extreme value occurrences caused by total radiative
569 contribution anomalies or component contribution anomalies in both hemispheres (Fig. S11), 52 % do
570 not exhibit extreme values in total radiative contribution due to the cancellation of contribution anomalies
571 between different components. For example, in the 70° S-80° S latitude zone (Fig. 4), the cloud
572 contribution in 2001 shows a positive anomaly and is the largest among the 21 years. However, since the
573 contribution from the surface and clear-sky atmospheric components are negative anomalies, the total
574 radiative contribution does not reach extreme levels.

575 In the latitude zones of 0°-30°, the variability of total radiation contribution is predominantly
576 influenced by the anomalies in cloud component contribution, confirming previous studies that attribute
577 tropical albedo variability primarily to cloud variability, especially associated with the El Niño-Southern
578 Oscillation (ENSO) phenomenon (Loeb et al., 2007; Jönsson and Bender, 2022). For instance, in the 0-
579 10° latitude zone of the SH, the extremely high anomaly in cloud contribution in 2015 can be attributed
580 to the exceptionally strong and prolonged El Niño event that occurred in the east-central equatorial
581 Pacific during 2015/2016 (Huang et al., 2016). The persistent anomalous ascending motions and large



582 amounts of water vapor in the east-central equatorial Pacific led to increased cloud formation (Avery et
583 al., 2017; Lim et al., 2017), resulting in higher reflected radiation from the cloud components.
584 Additionally, the significant positive anomaly in clear-sky atmospheric contribution can be linked to
585 smoke pollution caused by extensive fires in equatorial Asia during September-October 2015 (Koplitz et
586 al., 2016). It is observed that the combination of positive extreme anomalies in cloud state parameters
587 (CVOD and CF) and clear-sky atmospheric parameters (TCWV and AOD) greatly influences the
588 reflected radiation contribution (Fig. S8). Although there are large negative anomalies in SM due to
589 extreme drought in the Amazon region (Jiménez-Muñoz et al., 2016), the impact of SM anomalies on the
590 surface component contribution anomaly is limited in this primarily oceanic latitude zone. In 2010, the
591 20° S-30° S region was affected by a strong La Niña event, leading to anomalously heavy precipitation
592 in Australia and South Africa and extreme positive cloud component contribution (Lim et al., 2016;
593 Shikwambana et al., 2023). It is noteworthy that during this strong La Niña event, both the 0°-10° S and
594 10° S-20° S latitude zones exhibit dominant cloud contribution anomalies, but with opposite signs. A
595 similar situation is observed during the 2015 El Niño event. This suggests the presence of some
596 complementary mechanism between different latitudinal zones for radiative anomalies caused by such
597 large and complex weather patterns. At 20° S-30° S, the extremely high reflected radiation contribution
598 in 2019 is primarily contributed by the clear-sky atmosphere, which may be linked to the significant
599 aerosol emissions from severe forest fires in Australia (Khaykin et al., 2020). This is supported by Fig.
600 S8, which indicates that the anomaly of AOD for this latitude zone in 2019 exceeds 140 % of its standard
601 deviation. In the 30° S-40° S latitude zone, although the total reflected radiation contribution is not
602 extreme in 2019, the clear-sky atmospheric contribution exhibits a positive extreme anomaly that is
603 counteracted by a negative contribution from the cloud component. This negative anomaly in cloud
604 contribution may be associated with the combined effects of a positive Indian Ocean Dipole (IOD) and
605 a central-Pacific El Niño on Australia during that period (Wang and Cai, 2020). At 70° S-90° S, while
606 the surface component contributes the most to the reflected radiation (Fig. 2), it does not contribute as
607 significantly to the variability of reflected radiation as the cloud component. Instead, in the 60°-70°
608 latitude zone of the SH, the variation in reflected radiation contribution is primarily influenced by the
609 surface component contribution, and its anomaly has an opposite effect to that of the cloud component
610 anomaly. This indicates that clouds moderately mitigate the impact of sea ice changes on the total



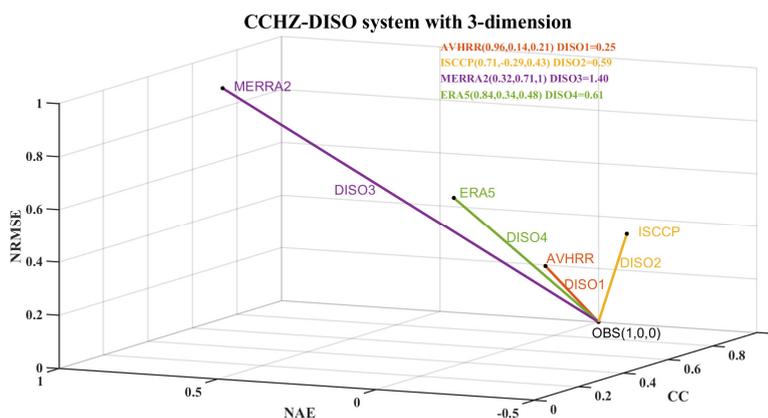
611 reflected radiation contribution. The rapid expansion of Antarctic sea ice prior to 2014 results in
612 anomalous ISC and a positive surface component contribution anomaly in this region during 2013-2014
613 (Riihelä et al., 2021).

614 The equatorial to 10° N region experiences strong negative anomalies in cloud state parameters (CF,
615 CVOD, LWP) in 2009 (Fig. S9), leading to an extreme anomaly in the cloud component contribution.
616 This is caused by a record-breaking warming in SST in the tropical North Atlantic starting in the summer
617 of 2009. This warming is a typical response to ENSO and is influenced by the negative phase of the
618 North Atlantic Oscillation (NAO) (Hu et al., 2011). In this region, 2016 is an extreme low year
619 characterized by a negative anomaly in cloud component contribution due to the presence of a strong
620 negative IOD and a weak La Niña (Lim and Hendon, 2017). This results in an abnormal decrease in
621 cloudiness in the equatorial central Pacific region. In the mid-latitudes of the NH (Fig. 5), the contribution
622 of the clear-sky atmosphere is more prominent due to the stronger influence of human activities in this
623 region, with sulfate aerosols being the dominant aerosol component (Diamond et al., 2022). In 2019, the
624 30° N-40° N latitude zone exhibits a negative anomaly in the contribution of the clear-sky atmospheric
625 component, primarily due to a significant reduction in atmospheric aerosols over much of east-central
626 China. This reduction is a result of emission reductions implemented during the COVID-19 outbreak as
627 part of epidemic control measures in China (Letu et al., 2023). In the Arctic, the contribution of the
628 surface component is minimal (Fig. 2), but it becomes the secondary dominant component in the variation
629 of the total radiative contribution after the cloud component. However, overall, the total radiative
630 contribution anomaly in the Arctic is still primarily influenced by the anomaly in the cloud component
631 contribution.

632 In conclusion, whether in low or high latitudes, and whether considering long-term perturbations or
633 extreme events, the impact of cloud variability on changes in the contribution of reflected radiation in
634 different latitudinal zones cannot be ignored. Therefore, accurate modeling of the cloud component is
635 crucial. It has been demonstrated that climate projections are sensitive to different parameterization
636 schemes for cloud radiation (Li and Le Treut, 1992; Li et al., 2022). If the cloud parameterization scheme
637 is not well-refined, the model will struggle to accurately simulate reflected radiation at the TOA, making
638 it challenging to explore the potential mechanisms behind the hemispheric symmetry of the PA.

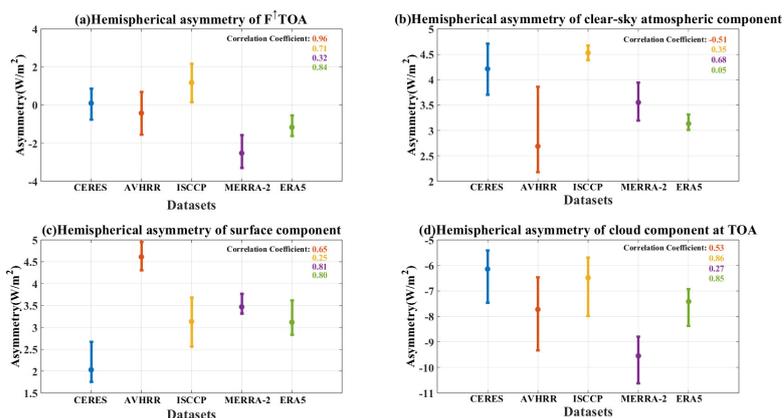


639 **3.3 Performance of different datasets on hemispheric asymmetry**



640

641 **Figure 6: CCHZ-DISO system with 3-dimension. The coordinate axis consists of three statistical indicators,**
 642 **correlation coefficient (CC) for x-axis, normalized absolute error (NAE) for y-axis, and normalized root mean**
 643 **square error (NRMSE) for z-axis. OBS indicates observations, here referred to as CERES EBAF.**



644

645 **Figure 7: Multi-year averages of hemispheric differences in (a) TOA SW upwelling radiation flux and its (b)**
 646 **clear-sky atmospheric, (c) surface, and (d) cloud components for the five datasets from Mar./2001-Feb./2016,**
 647 **with the maximum annual average difference for the dataset at the top of the error bars and the minimum at**
 648 **the bottom of the error bars. Blue for CERES EBAF, orange for Cloud_cci AVHRR, yellow for ISCCP, purple**
 649 **for MERRA-2, and green for ERA5. The numbers in the upper right corner are the correlation coefficients**
 650 **of time series for hemispheric differences of the different datasets with CERES EBAF.**

651 As mentioned earlier, AVHRR, ISCCP, MERRA-2, and ERA5 provide longer-term TOA reflected
 652 SW flux data compared to CERES EBAF. If these datasets can reproduce the observed hemispheric
 653 symmetry of reflected radiation captured by CERES, it would greatly assist in identifying the underlying



654 mechanism responsible for the hemispheric symmetry of PA using data spanning longer time periods.
655 Therefore, in order to comprehensively assess the performance of each dataset, Figure 6 presents three-
656 dimensional results based on CERES EBAF data using the CCHZ-DISO data evaluation system.
657 Different assessment metrics yield different results (Table S1), highlighting the importance of selecting
658 appropriate assessment metrics based on the specific research needs to obtain the most suitable dataset.
659 In general, the results indicate that AVHRR has the closest DISO value to CERES (DISO1=0.25) and
660 exhibits the best performance in terms of hemispheric symmetry. It is followed by ISCCP (DISO2=0.59)
661 and ERA5 (DISO4=0.61), while MERRA-2 performs the worst (DISO3=1.40). It should be noted that
662 the inclusion of spatial correlation coefficient in the DISO system did not significantly alter the results
663 (Table S1c), so the three recommended metrics (CC, NAE, and NRMSE) are still used. Additionally,
664 AVHRR demonstrates the highest correlation coefficient (0.96) with CERES for the time series of
665 hemispheric differences in total reflected radiation, and their multi-year averages are the closest (Fig. 7a).
666 Lim et al. (2021) have shown that ERA5 exhibits good agreement with CERES in simulating the inter-
667 annual variation of global TOA SW reflected radiation. Moreover, ERA5 and ISCCP display good
668 correlations with CERES (0.84 and 0.71, respectively), despite slight underestimation and overestimation
669 of NH reflected radiation. Among all the datasets, MERRA-2 exhibits the poorest performance in terms
670 of hemispheric symmetry, which may be primarily influenced by cloud cover bias (Lim et al., 2021).

671 To better investigate the performance of hemispheric symmetry in the reflected solar radiation
672 across different datasets, Figure S12 illustrates the variation in multi-year average hemispheric
673 differences of the reflected radiation and its components over time. We aim to determine the timescale
674 suitable for studying the hemispheric symmetry of PA using these datasets. Since previous studies lacked
675 a clear definition of PA's hemispheric symmetry, we will discuss this issue here. Voigt et al. (2013)
676 conducted a random division of the Earth into two halves to assess whether these random pairs exhibited
677 hemispheric symmetry in reflected solar radiation. The results revealed that only 3 % of the random pairs
678 demonstrated a hemispheric difference in reflected radiation smaller than 0.1 W m^{-2} , as measured by
679 CERES-EBAF. Furthermore, even when this criterion was extended tenfold (1 W m^{-2}), only 31% of the
680 random pairs satisfied the hemispheric symmetry requirement. Stephens et al. (2015) noted that the multi-
681 year average hemispheric difference in reflected solar radiation between the NH and SH is less than 0.2
682 W m^{-2} , suggesting this as an indicator of hemispheric symmetry. Here, when we use a symmetry criterion



683 of 0.1 W m^{-2} , CERES achieves hemispheric symmetry of reflected radiation on a 16-year annual mean
684 scale, while none of the other datasets do. When we expand this symmetry criterion to 0.2 W m^{-2} , the
685 symmetry study application of CERES is around 9-year scale, and other datasets remain inapplicable.
686 When held to a more conservative standard of 1 W m^{-2} , CERES achieves hemispheric symmetry every
687 year, and AVHRR achieves it on scale of more than two years. Interestingly, the ISCCP exhibits
688 increasing hemispheric asymmetry with longer durations, only declining after a 13-year average. ERA5
689 also displays a similar but more moderate increase in hemispheric asymmetry. In addition, in order to
690 have a more rigorous standard, we would like to take the uncertainty of the instrumental measurements
691 into account. If the difference between the solar radiation reflected from the NH and SH is within the
692 uncertainty of the measurement, it is considered as hemispherical symmetry (Diamond et al., 2022). The
693 regional averaged monthly mean uncertainty of the reflected SW radiation at the TOA of the CERES
694 EBAF is 2.5 W m^{-2} (Loeb et al., 2018b). AVHRR, ISCCP, MERRA-2 and ERA5 have monthly regional
695 mean biases from the CERES of 3.3 W m^{-2} , 4.8 W m^{-2} , 5.9 W m^{-2} and -1.9 W m^{-2} , respectively, which
696 will be used to calculate their uncertainties. Here we follow the method of Jönsson and Bender (2022) to
697 calculate the uncertainty of hemispheric difference of reflected solar radiation flux, noting that only rough
698 calculations have been made due to the unavailability of uncertainties at different grid points around the
699 globe. Uncertainty in the time-mean over the N-month period is scaled by a factor of $N^{-1/2}$. Then we can
700 obtain a time series of the uncertainty in the hemispherical differences of reflected radiation for each
701 dataset (Fig. S12). It is clear that as time grows, the range of uncertainty shrinks. Note that if the solid
702 black line falls within the shaded area, it indicates that the total reflected radiation exhibits credible
703 hemispheric symmetry within the given uncertainty. The hemispherical difference of the total reflected
704 radiation from CERES remains well within its uncertainty range (not shown). Similarly, AVHRR
705 demonstrates good agreement with uncertainty over a 14-year timescale. On the other hand, ISCCP only
706 maintains this agreement on timescales up to 5 years. The reanalyzed datasets significantly deviate from
707 their respective uncertainty ranges. In summary, AVHRR shows better consistency with CERES
708 regarding the hemispheric symmetry of the total reflected radiation.

709 Some datasets perform well in total reflected radiation symmetry, it doesn't mean that they also can
710 reasonably simulate the components. Here, we further decompose the TOA total reflected radiation of
711 these datasets into clear- sky atmosphere, surface and cloud components, and compare the performance



712 of the five datasets on hemispheric differences (Fig. 7). Although AVHRR exhibits well symmetry in the
713 total reflected radiation, it shows considerable deviation from CERES in the hemispherical difference of
714 three components. AVHRR exhibits a brighter SH due to its clear underestimation of the clear-sky
715 atmospheric component contribution in the NH and an overestimation of the cloud component
716 contribution in the SH. Particularly, it significantly overestimates the surface component of the NH.
717 Interestingly, AVHRR even demonstrates a negative correlation (-0.51) with CERES concerning
718 hemispheric differences in the clear-sky atmospheric component, and the data itself displays a high
719 degree of dispersion. This can be attributed, in part, to the limitations of the current AVHRR version
720 regarding aerosols. The aerosol optical thickness is set at 0.05, which is considered an underestimation
721 for high aerosol load situations (Stengel et al., 2020). Furthermore, AVHRR notably overestimates the
722 surface component in the NH. The PVIR mentions that the Cloud_cci dataset exhibits higher biases in
723 TOA upwelling shortwave flux compared to CERES in regions with low vegetation and typically high
724 surface albedo. Additionally, AVHRR demonstrates a significant deviation from CERES in the
725 hemispheric differences of the cloud component. Stengel et al. (2020) highlighted that AVHRR PMv3
726 shows a greater bias in identifying liquid clouds and reducing ice water paths compared to v2. In terms
727 of ISCCP, the multi-year means of the hemispheric differences for all three components are closest to
728 CERES among the datasets. However, the time series of annual means for the hemispheric difference in
729 the clear-sky atmospheric component and the surface component show poor correlation with CERES.
730 On the other hand, ISCCP performs well in evaluating the hemispheric differences in the cloud
731 component, showing close agreement and a strong correlation with CERES (0.86). Previous studies have
732 indicated that ISCCP employs a visible adjustment to correct for emission through thin cirrus clouds,
733 which enhances the accuracy of cloud top pressure retrievals for single-layer cirrus clouds (Marchand et
734 al., 2010). Additionally, the incorporation of the Max Planck Institute Aerosol Climatology (MAC) in the
735 treatment of stratospheric and tropospheric aerosols within the ISCCP-H series helps reduce the
736 misidentification of aerosols as clouds (Young et al., 2018). However, the aerosol input dataset MAC-v2,
737 despite improving cloud simulation, suffers from significant errors and uncertainties (further details in
738 the 'ISCCP-FH Radiative Flux Profile Product C-ATBD'). This may explain the lack of correspondence
739 between the hemispheric differences in the clear-sky atmospheric components of ISCCP and CERES.
740 Nevertheless, ISCCP still exhibits biases in cloud retrieval. For instance, Boudala and Milbrandt (2021)



741 discovered that ISCCP data overestimates cloud cover in both hemispheres between approximately 40°
742 and 60° latitude, particularly in North America and Europe, but significantly underestimates cloud cover
743 in the tropics and high latitudes in both hemispheres. Furthermore, ISCCP also overestimates the total
744 cloud fraction of TP compared to other space-based lidars (Zhao et al., 2023). The two reanalysis datasets
745 exhibit different performance in simulating hemispheric differences in the three components. MERRA-
746 2 poorly models the hemispheric difference in the cloud component, whereas ERA5 demonstrates good
747 agreement and correlation with CERES. Hinkelman (2019) pointed out that the discrepancy in all-sky
748 reflected SW radiation flux at TOA between MERRA-2 and EBAF is attributed to differences in cloud
749 variables such as cloud fraction or optical depth. In MERRA-2, there is an excessive presence of clouds
750 over tropical oceans and a slight underestimation of clouds in the oceanic stratocumulus region, as well
751 as an overestimation of clouds in the Southern Ocean (Hinkelman, 2019). It has been shown that
752 MERRA-2 consistently underestimates total cloud cover across almost all areas of the TP at all times
753 (Deng et al., 2023) and altitudes (Zhao et al., 2023). This bias may stem from a flaw in the cloud
754 parameterization within the reanalysis assimilation model (Dolinar et al., 2016; Li et al., 2017). When
755 assessing the cloud properties of reanalysis data over East Asia, Yao et al. (2020) found that ERA5 and
756 MERRA-2 generally overestimate liquid clouds, with MERRA-2 also overestimating ice clouds over the
757 cyclone center. Nevertheless, ERA5 still exhibits good correlation with CERES, except for the
758 hemispheric difference in the clear-sky atmospheric component. Li et al. (2023) demonstrated that the
759 deviation of ERA5's surface solar radiation products from observed values increases with higher aerosol
760 loading, indicating that aerosols have a significant impact on the accuracy of ERA5's radiation products.
761 Furthermore, the cumulative annual mean time series of hemispheric differences in the cloud component
762 for ISCCP and ERA5 display similar variations to CERES, with ISCCP exhibiting a smaller bias.
763 However, the cumulative annual mean time series of hemispheric differences in the surface component
764 for ISCCP differs from all other datasets, as its hemispheric difference increases after the 5-year average
765 but ultimately converges to a constant at longer time scales as others. The cumulative annual mean of
766 hemispheric differences in the clear-sky atmospheric component for AVHRR, ISCCP, and MERRA-2
767 exhibits similar characteristics to CERES, with a pronounced change, illustrating the irregularity of
768 human activities. In contrast, ERA5 shows a consistent decline.

769 On the whole, if the focus is solely on studying the hemispheric symmetry of total reflected radiation,



770 AVHRR can be chosen, although it exhibits poor performance in simulating the hemispheric differences
771 of the components. Conversely, ISCCP can be utilized for studying the hemispheric asymmetry in the
772 cloud component of reflected radiation. Generally, there is scope for improvement in simulating the
773 components of reflected radiation in these datasets. Further research and algorithmic enhancements may
774 be required to address these issues and enhance the usability and accuracy of these datasets in studies
775 related to hemispheric symmetry.

776 **4 Discussion and Summary**

777 The hemispheric symmetry of the PA is a powerful feature of the Earth system, and the mechanisms
778 by which it is currently maintained remain unclear, posing a great challenge for improving the simulation
779 of climate models. Numerous scholars have proposed many possible compensatory mechanisms, and
780 many of the different mechanisms are not only limited by latitude but also have seasonal characteristics,
781 and if we resolve the energy down to monthly scales and latitudinal zones, we can capture more variations
782 in the mechanisms. Accordingly based on the TOA and surface radiative flux data from CERES EBAF
783 during March 2001 to February 2021, we build on the original interannual variability and further explore
784 the contribution and monthly variability of different latitudinal zones as well as the variability of the
785 different components across latitudinal zones, and moreover the manifestation of hemispheric symmetry
786 for the different datasets. The results are as follows:

787 (1) The total reflected SW radiation at the TOA shows a clear hemispheric symmetry in the long-
788 term trend of the annual and the multi-year average. The annual mean decreasing trend in the NH is
789 synergistically influenced by changes in Arctic snow and ice, reduction in anthropogenic aerosols at mid-
790 latitudes, and a decrease in low cloud cover in the Eastern Pacific. In contrast, the decreasing trend in the
791 SH is mainly due to the cloud component. The cumulative year-to-year averaging of the hemispheric
792 differences in reflected radiation and its components tend to be constant over time, yet the clear-sky
793 atmospheric component displays great variability owing to human activities. More cloud component
794 contribution from 0° N-10° N and more clear-sky atmospheric and surface component contribution from
795 10° N-60° N compared to the corresponding latitude zones in SH, result in the dominance of NH reflected
796 radiation at 0°-40°. However, greater surface component contribution from higher latitude zones in the
797 SH and significantly higher cloud contribution from 10° S-70° S compensate for this, resulting in the



798 reflected radiation of SH leading that of the NH in 50°-90°. Within the latitudinal zones from the equator
799 to 40°, the annual cycle of reflected radiation contribution is distinguished by a high in fall/winter and a
800 low in spring/summer, with higher contribution in the NH than in the SH. Nonetheless, beyond 50°, this
801 pattern is reversed entirely. This feature of the annual cycle in various latitudes is linked to that of the
802 incident radiation and is mainly contributed by the cloud and clear-sky atmospheric component at low-
803 mid latitudes, and by the surface component at high latitudes. At high latitudes, the peak of the surface
804 component contribution in the NH is one month earlier than in the SH due to surface albedo and Arctic
805 amplification effects. From the perspective of hemispheric difference, the annual cycle of total reflected
806 radiation and its interannual variability are almost drive from those of cloud component contribution in
807 mid-low latitudes, and from surface and cloud contributions in high latitude zones.

808 (2) In extreme anomaly years across all latitudinal zones, the primary factor is the cloud component
809 contribution, followed by the surface component, and then the clear-sky atmospheric component. Surface
810 component anomalies play a larger role in extreme events in the SH, particularly in the 60° S-70° S. It is
811 common to observe cancellation of contribution anomalies between different components and latitudinal
812 zones during extreme events. The effect of cloud variability on the variation of reflected radiative
813 contributions of different latitudinal zones is not negligible at both low and high latitudes, for both long-
814 term perturbations and extreme events, and is evidently influenced by large-scale weather patterns such
815 as ENSO and PDO.

816 (3) According to the DDZJ-DISO assessment system, AVHRR performs best in terms of
817 hemispheric symmetry, followed by ISCCP and ERA5, and the worst is MERRA-2. Under different
818 symmetry criteria (0.1 W m⁻², 0.2 W m⁻², 1 W m⁻² and uncertainties of different datasets), the applicability
819 of different datasets to hemispheric symmetry of PA studies varies. AVHRR correlates best with CERES
820 for the time series of hemispheric differences of total reflected radiation, the multi-year mean is closest,
821 and the hemispheric symmetry of the PA can be studied within its uncertainty on 14-year timescale.
822 However, AVHRR does not adequately capture the hemispheric differences of individual components.
823 ERA5 and ISCCP simulate hemispheric differences of total reflected radiation with about the same
824 performance, while ERA5 itself is more stable. If one only wants to study cloud component, consider
825 using ISCCP.

826 Based on long-term satellite observations, our study and previous research have confirmed a clear



827 decreasing trend in solar radiation reflected back into space in both hemispheres over the past two
828 decades (Loeb et al., 2020; Stephens et al., 2022). This trend is particularly significant as the increase in
829 net solar energy absorption by the Earth outweighs the increase in outgoing thermal infrared radiation.
830 Consequently, it leads to a rise in global average temperature, exacerbating the Earth's energy imbalance,
831 global warming, sea-level rise, changes in the climate system, and the melting of glaciers and permafrost
832 (Loeb et al., 2021b). Given the profound impact of these changes on the climate system, it is crucial to
833 pay closer attention to the future evolution of PA and its symmetry. Although climate models persistently
834 exhibit biases in simulating the average state of albedo symmetry from CMIP3 to CMIP6 (Crueger et al.,
835 2023), they remain a powerful tool for generating hypotheses about the unexplained observed albedo
836 symmetry (Rugenstein and Hakuba, 2023) and projecting future evolutions and potential influencing
837 mechanisms. For example, Rugenstein and Hakuba (2023) examined the response of modeled surface
838 temperature and PA to CO₂ and found an increasing difference in surface warming between the two
839 hemispheres under stronger carbon dioxide forcing and weaker aerosol forcing. They also proposed that
840 the warmer hemisphere will become darker, suggesting a potential asymmetry in albedo in the coming
841 decades. On the other hand, Diamond et al. (2022) focused on changes in clear-sky hemispheric
842 asymmetry under different emission scenarios simulated by their model. Their results indicated a
843 significant shift in clear-sky albedo asymmetry throughout this century under both high and low emission
844 scenarios, primarily driven by anthropogenic aerosol emissions and cryosphere changes. Furthermore,
845 Jönsson and Bender (2023) investigated the evolution of hemispheric albedo differences following a
846 sudden quadrupling of CO₂ concentration using CMIP6 coupled model simulations. They found that the
847 initial albedo reduction in the NH may be partly compensated by a reduction in extratropical cloudiness
848 in the SH on a much longer timescale which can be referred to as a mechanism of trans-hemispheric
849 communication. They also highlighted that if PA maintains hemispheric symmetry, compensating for
850 cloud variations will have uncertain but significant effects on Earth's energy balance and hydrologic
851 cycle. However, whether the hemispheric symmetry of PA can be sustained indefinitely remains an open
852 question. Therefore, it is essential to focus on investigating additional potential mechanisms of
853 hemispheric albedo symmetry and future projections using model ensembles, along with observational
854 constraints.

855



856 **Data availability.** The CERES_EBAF_Ed4.2 and CERES-SSF1deg products are publicly available
857 through the NASA Langley Research Center CERES ordering tool at <https://ceres.larc.nasa.gov/data/>.
858 The ESA Cloud-cci version 3 products, AVHRR-PMv3 for this research are included in the paper: Stengel
859 et al. (2020), or obtained through
860 https://public.satproj.klima.dwd.de/data/ESA_Cloud_CCI/CLD_PRODUCTS/v3.0/L3C/. The ISCCP-
861 FH data are available from the following website: <https://isccp.giss.nasa.gov/pub/flux-fl/>. The MERRA-
862 2 datasets used in this study are available from the following websites:
863 <https://doi.org/10.5067/OU3HJDS973O0>, <https://doi.org/10.5067/OU3HJDS973O0> and
864 <https://doi.org/10.5067/FH9A0MLJPC7N>. The ERA5 monthly averaged data on single levels from 1940
865 to present are available from Climate Data Store (CDS) of
866 [https://cds.climate.copernicus.eu/cdsapp#!/dataset/reanalysis-era5-single-levels-monthly-](https://cds.climate.copernicus.eu/cdsapp#!/dataset/reanalysis-era5-single-levels-monthly-means?tab=overview)
867 [means?tab=overview](https://cds.climate.copernicus.eu/cdsapp#!/dataset/reanalysis-era5-single-levels-monthly-means?tab=overview).

868

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871 DW and LZ modified the paper and provided suggestions for this study. All authors contributed to the
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873

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890

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