



1 **Influence of Cloud Retrieval Errors Due to Three Dimensional Radiative Effects on**
2 **Calculations of Broadband Cloud Radiative Effect**

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10



11

Abstract

12

13 We investigate how cloud retrieval errors due to the three-dimensional (3D) radiative effects affect
14 broadband cloud radiative effects (CRE). A framework based on the combination of large eddy simulations
15 (LES) and radiative transfer (RT) models was developed to simulate both one-dimensional (1D) and 3D
16 radiance, and shortwave (SW) broadband fluxes. Results show that the broadband SW fluxes reflected at
17 top-of-the-domain, transmitted at the surface, and absorbed in the atmosphere, computed from the
18 cloud retrievals using 1D-RT (called "1D-RT+retrieved clouds") can provide reasonable broadband
19 radiative energy estimates in comparison with those derived from the true cloud fields using 1D-RT (called
20 "1D-RT+true clouds"). The difference between these 1D-RT simulated fluxes (1D-RT+retrieved clouds
21 simulations, 1D-RT+true clouds simulations) and the benchmark 3D-RT simulations from the true cloud
22 field (called "3D-RT+true clouds"), depends primarily on the horizontal transport of photons in 3D-RT,
23 whose characteristics vary with the Sun's geometry. When the solar zenith angle (SZA) is 5° , the domain-
24 averaged fluxes simulated based on the 1D-RT+retrieved clouds are in excellent agreement with the 3D-
25 RT+true clouds results, all within 7% relative CRE bias. When the SZA is 60° , the differences between the
26 results from the 1D-RT+retrieved clouds and 3D-RT+true clouds are determined by how the cloud side-
27 illumination and shadowing effects offset each other in the radiance, retrieval, and broadband fluxes. This
28 study suggests that although the cloud property retrievals based on the 1D-RT theory may be biased due
29 to the 3D radiative effects, they still provide an observational basis for the estimation of broadband fluxes.

30



31 1. Introduction

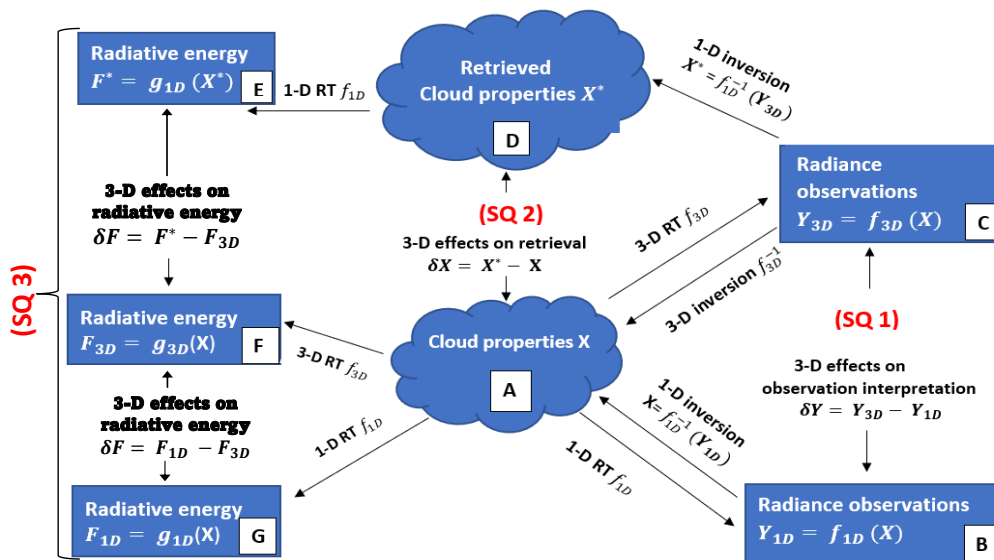
32 Covering about 60-70% of the Earth's surface [Rossow and Schiffer, 1999; Vardavas and Taylor,
33 2011], clouds play a very important role in the Earth's climate system. Clouds can cool the Earth by
34 reflecting shortwave (SW) solar radiative flux back to space and at the same time warm the Earth by
35 retaining the outgoing longwave (LW) infrared radiative flux at the top of the atmosphere (TOA), known
36 as the cloud radiative effects (CRE). The annual global average TOA CRE is approximately -50 Wm^{-2} at
37 SW and 30 Wm^{-2} at LW, resulting in a net CRE of about -20 Wm^{-2} [Intergovernmental Panel on Climate
38 Change (IPCC) Fifth Assessment Report Chapter 7]. These strong CRE show that clouds greatly affect the
39 Earth's energy budget [Ramanathan et al., 1989; Kiehl and Trenberth, 1997; Trenberth et al., 2009]. The
40 CRE of clouds is largely determined by the optical and microphysical properties of clouds including the
41 cloud optical thickness (τ), cloud droplet effective radius (r_e), and cloud liquid water path (LWP). Thus,
42 continuous measurements of these cloud properties from regional to global scales are critical to better
43 understand the role of clouds in the climate systems. Currently, satellite based remote sensing is the only
44 way to make such observations. Remotely "retrieved" cloud properties based on these satellite
45 observations are often used to derive the radiative effects of clouds [e.g., Wielicki et al., 1996; Platnick
46 et al., 2003; Loeb and Manalo-Smith, 2005; Oreopoulos et al., 2016] and evaluate the simulations of Earth
47 System Models (ESMs) [Kay et al., 2012; Nam et al., 2012; Song et al., 2018].

48 A commonly used retrieval technique in passive satellite remote sensing is the bi-spectral retrieval
49 method first developed by Nakajima and King [1990]. It retrieves τ and r_e simultaneously from a pair of
50 total reflectance measurements, one from the non-absorbing visible or near infrared (VNIR) band (e.g.,
51 $0.66 \mu\text{m}$) and the other from the moderately absorbing short-wave infrared (SWIR) band (e.g., $2.13 \mu\text{m}$).
52 Since clouds in reality have three-dimensional (3D) structures, the simulation of radiative transfer (RT) in
53 clouds should ideally consider the transport of radiation in both vertical and horizontal directions (referred
54 to as "3D RT"). Unfortunately, the computational cost for 3D RT is extremely high. As a result, the
55 operational bi-spectral cloud retrievals are almost exclusively based on the one-dimensional (1D) RT
56 theory that considers only the vertical and ignores the net horizontal transport of radiation. The radiative
57 properties of clouds under 3D RT are substantially different from those under 1D RT. This is known as the
58 3D radiative effects and can lead to substantial biases in the cloud property retrievals based on 1-D RT
59 [Várnai et al., 2001; Marshak et al., 2006; Zhang et al. 2012; Zhang et al. 2016]. Although recent efforts
60 have been made to employ machine learning techniques to retrieve cloud properties based on 3D RT
61 theory [Okamura et al., 2017; Masuda et al., 2019], these machine-learning based algorithms are still in
62 their infancy and far from being used in operational algorithms.

63 Many previous studies have investigated the 3D radiative effects on satellite radiance
64 observations and cloud property retrievals. For example, Welch and Wielicki [1984] used some "toy" cloud
65 fields (e.g., cubic, and cylindrical) to illustrate the impact of side-illuminating and mutual shadowing on
66 cloud albedo. Várnai and Davis [1999] and Várnai [2000] elucidated several 3D RT mechanisms, e.g.,
67 upward/downward trapping/escaping, that can result in significant differences between 1D and 3D cloud
68 albedo. Davis and Marshak [2001] pointed out that the channeling effect in 3D RT can smoothen out the
69 small-scale cloud variations and lead to the reduction of cloud brightness at cloud edges. Marshak et al.
70 [2006] explained how the radiance biases due to 3D radiative effects can lead to τ and r_e retrieval biases
71 in MODIS (Moderate Resolution Imaging Spectroradiometer) cloud products. This study is built upon these
72 classic papers but has a different objective.



73 Here, we investigate an important question: Do cloud property retrievals based on 1D RT, which
 74 are potentially biased due to the 3D radiative effects, still provide an observational basis to estimate the
 75 broadband CRE? This is an important question because as mentioned above, operational cloud retrieval
 76 products from, for example MODIS, are frequently used for CRE estimation and ESM evaluation. However,
 77 to our best knowledge, the impacts of retrieval bias due to the 3D radiative effects on such applications
 78 have never been examined systematically in previous studies. To better explain our objective and the
 79 difference of this study from many previous ones on the 3D radiative effects, we need to introduce a
 80 framework illustrated in **Fig. 1**. As conceptually illustrated in **Fig. 1**, the observed radiances are inherently
 81 3D (i.e., from Box A to C) because the RT in nature is 3D. However, when 1D RT theory is used to interpret
 82 the observations, we get the “retrieved cloud properties” in Box D that can be significantly different from
 83 the “true” cloud properties in Box A. Although the retrieved cloud properties are often biased due to the
 84 3D radiative effects, they are still widely used to compute the radiative fluxes by clouds (i.e., from Box D
 85 to E) using 1D RT and the results are often used for studying the climatic effects of clouds [e.g., Kato et al.,
 86 2011; Zelinka et al., 2012; Oreopoulos et al., 2016]. In contrast, the “true” radiative fluxes in nature are
 87 also 3D (i.e., from Box A to F). A few recent studies have computed and compared the 1D and 3D radiative
 88 fluxes and heating rates by clouds. For example, Barker et al. [2011, 2012] and more recently Okata et al.
 89 [2017] compared the 1D and 3D SW fluxes computed based on the constructed A-Train cloud scenes at
 90 the TOA and surface. The main difference between their study and this current work is as follows: They
 91 compared the 3D (i.e., Box F in **Fig. 1**) with the 1D broadband fluxes (i.e., Box G in **Fig. 1**) both computed
 92 from the “true” clouds. In contrast, we argue that the “true” clouds are not known in practice and
 93 therefore it is more reasonable to compare the 3D flux (i.e., Box F in **Fig. 1**) with the 1D flux computed
 94 from the “retrieved cloud properties” (i.e., Box E in **Fig. 1**).



95
 96

Fig. 1. Conceptual framework to understand the study.



97 The above analysis of **Fig. 1** reveals a big uncertainty in the current studies of CRE, i.e., we know the
98 retrieved cloud properties are biased due to the 3D radiative effects but still have to use them for the CRE
99 computations. This raises a highly important question: Do cloud property retrievals based on 1D RT, which
100 are potentially biased due to 3D radiative effects, still provide an observational basis to estimate the
101 broadband CRE? This question motivates this study and will focus on the three important scientific
102 questions (SQs) in the framework illustrated in **Fig. 1**:

- 103 • SQ1: How does the radiance simulated based on 3D RT model compare with the 1D results for
104 different types of clouds at different illuminating-viewing geometries? (i.e., Comparing Box C to B
105 in **Fig. 1**).
- 106 • SQ 2: How does the “retrieved cloud properties”, e.g., τ and r_e , derived based on the 3D radiance
107 field using 1D RT, compare to the “true” cloud properties? (i.e., Comparing Box D to A in **Fig. 1**).
- 108 • SQ 3: How are the broadband SW radiative fluxes derived from the retrieved cloud properties
109 (using 1D RT) different from the “true” radiative fluxes (computed from the “true” cloud fields
110 using 3D RT)? And how does this result compare with the difference between the broadband SW
111 radiative fluxes computed from the “true” cloud properties using 1D RT and those computed from
112 the “true” cloud properties but using 3D RT? (i.e., Comparing Box F to E and G in **Fig. 1**).

113 The paper’s remaining structure is arranged as follows: Section 2 briefly describes the data and theory for
114 the study. Section 3 presents and discusses results on how the 3D radiative effects influences the radiance
115 fields, cloud property retrievals and broadband radiative flux. The summary and conclusion are given in
116 Sect. 4.

117 **2. Data and Theory**

118 **2.1. Cloud field data set**

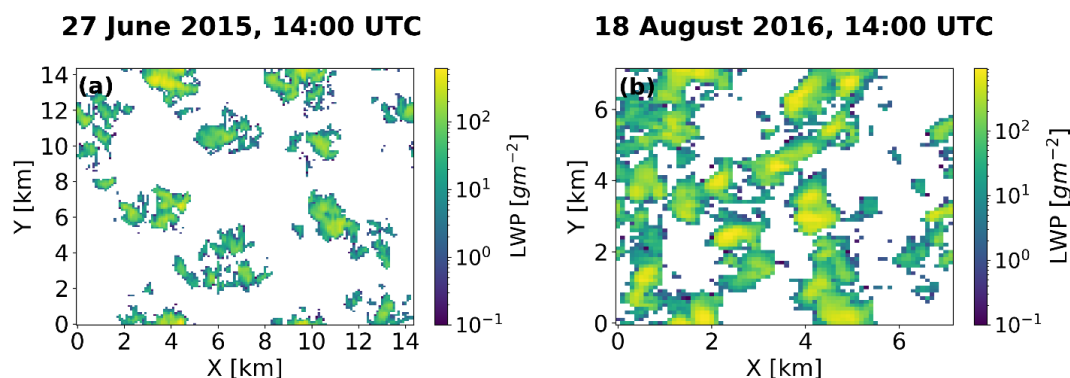
119 A great challenge facing 3D radiative effects studies is that the “true” clouds are always obscured by
120 the 3D radiative effects which are inevitable in real observations. To overcome this challenge, many
121 previous studies [e.g., Zhang et al., 2012; Miller et al., 2017; Rajapakshe and Zhang, 2020] have used
122 synthetic cloud fields and RT simulations to mimic the observation-retrieval process and study the 3D
123 radiative effects. Building on these previous studies, we adopt the same state-of-the-art satellite retrieval
124 simulator by Zhang et al. [2012] and added a broadband flux computation function to study the 3D
125 radiative effects and its impact on broadband SW radiative flux. As described in Zhang et al. [2012] and
126 illustrated in **Fig. 1**, the framework consists of three major components: 1) Synthetic cloud fields; 2) RT
127 models (for radiance and broadband flux simulations); 3) cloud property (e.g., τ and r_e) retrieval
128 simulator. Similar to Zhang et al. [2012], the synthetic cloud fields utilized in this study are based on Large-
129 --Eddy Simulations (LES) cloud fields.

130 Since the 3D radiative effects on overcast clouds are minimal, two cloud fields of low and
131 intermediate cloud fractions have been selected as a case study to illustrate the framework explained in
132 Sect. 1. The selected cloud fields were from the LES Atmospheric Radiation Measurement (ARM)
133 Symbiotic Simulation and Observation (LASSO) Activity, conducted in the ARM Southern Great Plain (SGP)
134 site located in Lamont, Oklahoma [Gustafson et al., 2020] (<https://www.arm.gov/capabilities/modeling/lasso/>). LASSO enhances ARM’s observations by using LES
135 modeling to provide contextual and self-consistent representation of the atmosphere surrounding the
136 ARM site. It also provides continuous observations from ground-based cloud and radiometric instruments
137



138 which is valuable for enhancing research on cloud-radiation interactions. For this study, the two snapshots
 139 of LASSO LES cloud field cases analyzed are: 14:00 UTC on June 27 June 2015, simulation ID=108 [ARM
 140 user facility, 2015] and the other at 14:00 UTC on 18 August 2016, simulation ID=113 [ARM user facility,
 141 2016]. For conciseness in this text, these snapshots will be referred to as “27 June” and “18 August”
 142 respectively. We chose to use these specific LASSO LES cloud fields data from the stated dates, because it
 143 represents typical shallow cumulus clouds, does not contain ice (to avoid the complexities dealing with
 144 ice microphysics) and has better diagnostic statistics compared to other LES data streams.

145



146

147 **Fig. 2.** Large-Eddy Simulation (LES) of cloud liquid water path (LWP) for 14:00 UTC, 27 June 2015 (a), and 14:00 UTC, 18 August
 148 2016 (b) at the ARM SGP atmospheric observatory. White areas are clear-sky regions where cloud liquid water path (LWP) =0.

149 The LASSO LES cloud fields are characterized by broken cloud pattern spatially distributed across
 150 the domain as seen in the LWP maps in **Fig. 2a and b** for the 27 June and 18 August cases, respectively.
 151 The 3D distribution of cloud liquid water content (LWC) was obtained from the LASSO cloud fields data
 152 and a two-moment bulk microphysics scheme by Morrison et al. [2008] (see their equation 5 in Sect. 2)
 153 was used to obtain the r_e associated with the corresponding LWC distribution. It is important to note that
 154 for this study, a cloudy column has been defined as a column with $LWP > 0$ (i.e., clear-sky regions have
 155 $LWP=0$). The cloud fields have different domain sizes and microphysics distribution, and the cloud cover
 156 for the 18 August cloud field (47.08%) is more than twice that of the 27 June cloud field (20.15%).
 157 Information about the cloud properties and the LES domain are summarized in **Table 1**.

158 **Table 1.** Cloud property characteristics for the LES cloud field cases. The mean cloud effective radius (r_e), mean cloud optical
 159 thickness (τ), and In-cloud liquid water path are from the average of the cloudy regions only. The columns from left to right are
 160 case name, cloud fraction, mean In-cloud liquid water path, cloud base height (CBH), cloud top height (CTH), mean r_e , mean τ ,
 161 grid spacing, and domain size, respectively.

Case name	CF (%)	Mean In-cloud LWP (gm^{-2})	CBH (km)	CTH (km)	Mean r_e (μm)	Mean τ	Grid spacing (m)	Domain size (km^3)
27 June 2015, 14:00 UTC	20.15	51.08	1.815	2.835	7.196	10.95	$\Delta x = \Delta y = 100, \Delta z = 30$	$14.4 \times 14.4 \times \sim 2.8$



18 August 2016, 14:00 UTC 47.08 127.67 0.945 2.355 8.020 23.24 $\Delta x = \Delta y = 100, \Delta z = 30$ 7.2 x 7.2 x ~2.4

162

163 2.2. Radiative Transfer Setup

164 We use the spherical harmonics discrete ordinate method (SHDOM) RT model developed by Evans
165 [Evans, 1998] to handle both 1D and 3D radiance computations. We have benchmarked the SHDOM
166 simulations against the results from our previous studies [Zhang et al., 2012; Miller et al., 2016].
167 Broadband SW radiative flux computations, both 1D and 3D, were performed with the Intercomparison
168 of 3D Radiation Codes (I3RC) Monte Carlo community model [Pincus and Evans, 2009], and atmospheric
169 gaseous absorption was incorporated via the SW Rapid Radiative Transfer Model (RRTM) correlated
170 k-distribution approach [Mlawer et al., 1997] which consists of 14 bands with spectral range from 0.2 to
171 12 μm (This coupled broadband radiative flux solver is hereafter known as the “I3RC+CKD” model).
172 Rayleigh scattering was included in the flux RT calculations, the background atmospheric profiles are taken
173 to be horizontally homogeneous throughout the domain and the profiles of atmospheric temperature,
174 pressure, ozone, air density and the water vapor profile utilized for the RT flux calculations were obtained
175 from the sounding data at the ARM SGP site on 27 June 2015. Ambient aerosols are neglected in the RT
176 calculations for simplicity. The 1D broadband RT flux calculations were performed with the same
177 I3RC+CKD model, by dividing the LES domain into individual columns and RT was calculated on each LES
178 column properties separately and independently.

179 The spectral cloud optical properties were calculated using Mie scattering theory and were
180 averaged over each RRTM spectral bands. The phase functions were represented using Legendre
181 coefficients with 35 log spaced effective radius spanning from 2 to 40 μm . The surface was assumed to be
182 Lambertian with surface spectral albedos obtained from ARM SGP site [Trishchenko et al., 2003] applied
183 for wavelength (λ) in the range $0.2 \leq \lambda \leq 2.5 \mu\text{m}$, while surface spectral albedo corresponding to a vegetative
184 covered surface [Zhuravleva et al., 2009] was utilized for $\lambda > 2.5 \mu\text{m}$. In the Monte Carlo calculations, 10^8
185 and 10^4 photons were initiated for calculations of the 3D broadband SW flux and the column-
186 independent 1D broadband SW flux, respectively. The radiative transfer calculations were implemented
187 for two solar zenith angles (SZAs), a high Sun case with SZA of 5° and a low Sun case with SZA of 60° . In
188 the broadband flux calculations, the downward flux at the top of the domain (TOD) corresponds to 1363
189 Wm^{-2} and 684.1Wm^{-2} for SZA 5° and 60° , respectively. Throughout this study, we choose a constant 0°
190 relative azimuth angle (RAA) and a constant 0° viewing zenith angle (VZA). The Double periodic horizontal
191 boundary conditions were applied for all the RT calculations, and all RT calculations have been conducted
192 at the native LES resolution of 100 m. Coarser spatial resolution will be applied in future studies.

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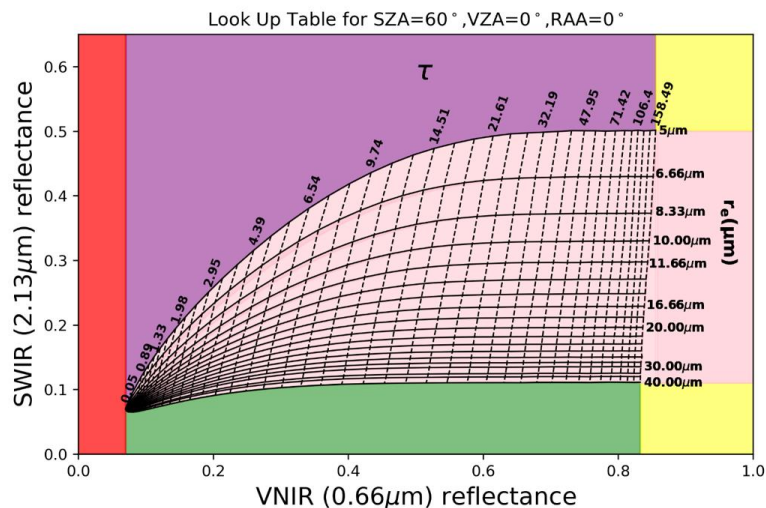
194 2.3. Bi-spectral retrieval method

195 As introduced in Sect. 1, one of the most widely used methods for retrieving the τ and r_e is the
196 bi-spectral retrieval technique proposed by Nakajima and King [1990]. This retrieval method uses passive
197 remote sensing measurements of the reflection function at a pair of wavelengths, one chosen from the
198 VNIR band where water has negligible absorption and therefore cloud absorption is more sensitive to the
199 τ and the other wavelength is chosen from the SWIR band where water has significant absorption and



200 therefore is more sensitive to the r_e . The bi-spectral retrieval method is solely based on the 1D RT theory
 201 to interpret the observed cloud reflectance. It is implemented using a precomputed Look up table (LUT)
 202 which consists of 1D reflectance function for different τ and r_e combinations at the required solar-view
 203 geometry (an example LUT is shown in **Fig. 3**). The observed cloud reflectance is then utilized as inputs to
 204 the LUT to simultaneously retrieve the τ and r_e via a two-dimensional (2D) interpolation between the
 205 observed cloud reflectance and the LUT grid. Notably, in the bi-spectral LUT regions with smaller τ , the
 206 retrieval uncertainty increases because the isolines of the LUT τ are less orthogonal and more tightly
 207 packed. This non-linearity in the LUT has high inhomogeneity consequences for cloud retrievals at the
 208 pixel level [Zhang et al., 2012, 2016]. In this study, the VNIR reflectance were measured at $0.66\ \mu\text{m}$
 209 (identical to the central wavelength of operational MODIS retrieval algorithm over a vegetative land
 210 surface), while the SWIR reflectance were measured at the $2.13\ \mu\text{m}$ wavelength. The LUT utilized for our
 211 bi-spectral retrievals have 19 effective radii spanning from 5 to $40\ \mu\text{m}$, and 43 log spaced τ values
 212 spanning from 0.05 to 158.48 . While a constant effective variance (v_e) value of 0.1 is used for consistency
 213 with all other RT simulations in this study. The surface albedo in both 0.66 and $2.13\ \mu\text{m}$ wavelengths for
 214 the LES radiance simulations and LUT RT calculations was 0.07 . This value is consistent with the surface
 215 albedo of similar spectral bands in the broadband SW flux computations.

216



217

218 **Fig. 3.** An example Nakajima and King bi-spectral Look up table (LUT) space. The solid lines are the reflectance function
 219 contours for fixed cloud effective radius (r_e), while the dashed lines are for fixed cloud optical thickness (τ). Surface is
 220 Lambertian with surface albedo= 0.07 . The solar zenith angle (SZA) is 60° , the view zenith angle (VZA) is 0° , and the relative
 221 azimuth angle (RAA) is 0° .

222 **2.4. Classification of failed and successful retrievals**

223 One major challenge in cloud property retrievals from satellite remote sensing instruments like
 224 MODIS, is the so called “failed retrievals”. A retrieval can be considered failed if there is no r_e and τ LUT
 225 grid combination to interpret the reflectance observation, or if there is no realistic cloud microphysics to
 226 explain the retrieved cloud property (e.g., a retrieved $r_e > 40\ \mu\text{m}$). These can be due to several factors,



227 such as the limits of the LUT, clouds overlapping effect, presence of partially cloudy pixels, extreme solar-
228 satellite viewing geometries, strategy used in cloud mask implementation and the optical characteristics
229 of the underlying surface. Potential causes and rate of occurrence of failed MODIS retrievals for marine
230 liquid phase clouds have been studied extensively [Cho et al., 2015]. In this study, we refer to MODIS cloud
231 property retrieval algorithm’s classification of failed retrievals [Platnick et al., 2016] and the study by Cho
232 et al. [2015], to classify a pixel as successful or failed retrieval as explained below:

- 233 1) For observations with both VNIR and SWIR reflectance observations within the LUT solution
234 space, the nearest interpolated τ and r_e values are retrieved (Pink area bounded by the LUT lines
235 in **Fig. 3**). If the observed VNIR reflectance exceed the upper limit of LUT τ but within the LUT r_e
236 solution range (extended pink area in **Fig. 3**), the nearest LUT r_e is retrieved and the maximum
237 LUT τ value ($\tau=158.48$) is assigned to the retrieval. These explained categories are classified as
238 “successful retrievals” for this study.
- 239 2) In other cases, for observations with VNIR reflectance within the LUT solution space but SWIR
240 reflectance above the LUT solution space (purple area in **Fig. 3**), the nearest τ values are retrieved
241 but the smallest LUT r_e value of $5 \mu\text{m}$ is assigned to the retrievals. This category of retrieval failure
242 is called “ r_e too small” failures. In cases where the VNIR reflectance observations are within the
243 LUT τ solution space, but the SWIR reflectance are below the LUT solution space (green area in
244 **Fig. 3**) the nearest τ values are retrieved but the largest LUT r_e value of $40 \mu\text{m}$ is assigned to the
245 retrieval. This category of retrieval failure is called the “ r_e too large” failures. In cases where the
246 observed VNIR reflectance is greater than the largest LUT τ value and the observed SWIR
247 reflectance is smaller than the largest LUT r_e (i.e., the lower yellow region in **Fig. 3**), the retrievals
248 are assigned the largest τ value ($\tau=158.48$) and the largest r_e value ($r_e=40 \mu\text{m}$). For observations
249 with VNIR reflectance greater than the largest LUT τ value and the SWIR reflectance greater than
250 the smallest LUT r_e value (i.e., the upper yellow region in **Fig. 3**), the retrievals are assigned the
251 largest τ value ($\tau=158.48$) and smallest r_e value ($r_e=5 \mu\text{m}$). Lastly, for observations with VNIR
252 reflectance below the minimum LUT τ (red area in **Fig. 3**), the r_e and τ retrievals are assigned fill
253 values (which are represented by $\tau = 0$ in our flux calculations). These explained categories are
254 called “ τ ” failures. The r_e too small, r_e too large and τ failure categories are collectively classified
255 as “failed retrievals” for this study.

256 2.5. Approach for radiative transfer simulation and result comparisons

258 To address the three SQs for our study (identified in Sect. 1), we performed a total of fourteen
259 experiments for each cloud field. The first four experiments were performed with the SHDOM model to
260 study the 3D radiative effects on radiance observation interpretation and address SQ1, it involves
261 comparing 1D and 3D RT radiance simulations (Box B vs C depicted in **Fig. 1**) for the high and low Sun
262 cases. The next four experiments involve comparing cloud property retrievals from 3D and 1D RT radiance
263 observations (Box B to A vs Box D in **Fig. 1**) for both high and low Sun, to examine the influence of the 3D
264 radiative effects on the retrieved cloud properties and address SQ2. These experiments were conducted
265 using radiance fields as inputs to the precomputed LUT described in Sect. 2.3. The last six experiments
266 were conducted to examine the impact of the 3D radiative effects on the broadband solar radiative flux
267 for both high and low Sun scenarios in the LES domains and address SQ3. These experiments involve the
268 calculation of the broadband solar radiative flux for each SZA within three distinct categories (**Table 2**).
269 The first set of flux was computed based on the “true” cloud field and 3D RT (i.e., Box F in **Fig. 1**) hereafter
270 referred to as the “3D-RT + true clouds” experiment. The second set was computed based on the



271 “retrieved cloud properties” using 1D RT (Box E in Fig. 1) hereafter referred to as the “1D-RT + retrieved
 272 clouds” experiment, and the third was computed based on the “true” cloud field and 1D RT (Box G in Fig.
 273 1) hereafter referred to as the “1D-RT + true clouds” experiment. It is important to note that in the 1D-
 274 RT + retrieved clouds experiment, the retrieved cloud properties (τ and r_e) are utilized to calculate the
 275 retrieved LWP (using retrieved $LWP \cong 2\tau\rho r_e/3$, where ρ is the density of liquid water; [Stephens, 1977;
 276 Liou, 1992]) which are then reconstructed into r_e and LWC distribution for each LES column while
 277 preserving the vertical structure of the original LES cloud field. 1D RT are then performed using the
 278 reconstructed retrieved clouds as inputs to obtain the 1D-RT + retrieved clouds experiment’s results. Note,
 279 unless otherwise stated, for this study, the successful and failed retrievals (as described in Sect. 2.4) have
 280 been used to represent the total population of cloudy pixels in the cloud property inputs for the 1D-RT +
 281 retrieved clouds experiment.

282 Table 2. Description of the Flux experiments

Name	Description
3D-RT + true clouds	Broadband SW flux computed based on “true” cloud field and 3D RT model
1D-RT + retrieved clouds	Broadband SW flux computed based on “retrieved cloud properties” and 1D RT model
1D-RT + true clouds	Broadband SW flux computed based on “true” cloud field and 1D RT model

283

284 The 1D-RT + true clouds experiment is identical to the 3D-RT + true clouds experiment except for
 285 the absence of the horizontal movement of photons between the LES grid columns. This enables us to
 286 determine the impact of neglecting the horizontal movement of photons on the broadband radiative
 287 fluxes. On the other hand, in reference to the 3D-RT + true clouds experiment, the 1D-RT + retrieved
 288 clouds experiment will not only help us to better understand the implications of neglecting the horizontal
 289 transport of photons but will also enable us to measure how biases in the retrieved cloud properties
 290 (which are affected by the 3D radiative effects) impact the broadband radiative fluxes.

291 In order to describe the impact of the 3D radiative effects on the radiance fields, retrieved cloud
 292 properties and broadband radiative flux, we first examine their effects across the LES domain and
 293 subsequently quantify their overall impact on the domain by computing the horizontally domain-averaged
 294 results and determine the absolute bias, hereafter referred to as “bias” for brevity and is defined as $\bar{y} - \bar{x}$
 295 , where \bar{y} denotes the domain-averaged result from the 3D RT quantity (e.g., Reflectance or flux), and \bar{x}
 296 denotes the domain-averaged result from the 1D RT quantity (e.g., Reflectance or flux).

297 To quantify the difference between the CRE computed from the benchmark 3D-RT + true clouds
 298 experiment and the CRE computed from the 1D-RT + true or 1D-RT + retrieved clouds experiment, we
 299 define a domain-scale quantity known as the relative cloud radiative effects (rCRE) bias as:

$$300 \quad rCRE \text{ bias} = \left(1 - \frac{CRE_{1D}}{CRE_{3D}}\right) \times 100 \quad (1)$$

301 Where CRE_{1D} is the CRE calculated from either the 1D-RT + true clouds or the 1D-RT + retrieved clouds
 302 experiment in units of Wm^{-2} and CRE_{3D} is the CRE calculated from the 3D-RT + true clouds experiment
 303 in units of Wm^{-2} . According to this definition, a rCRE bias of 0% would indicate that there is no bias



304 between the CRE computed from the 1D-RT + true clouds or 1D-RT + retrieved clouds experiments and
305 the CRE computed from the 3D-RT + true clouds experiment (i.e., the CRE computed from the 1D-RT +
306 true or 1D-RT + retrieved clouds experiment is equivalent to the CRE computed from the 3D-RT + true
307 clouds experiment), while a positive rCRE bias value greater than 0% would quantify the percentage by
308 which the CRE computed from the 1D-RT + true clouds or 1D-RT + retrieved clouds experiment is lesser
309 than the CRE computed from the 3D-RT + true clouds experiment and thus indicate that the 1D experiment
310 underestimate the CRE relative to the 3D-RT + true clouds experiment. Also, a negative rCRE bias value
311 less than 0% would quantify the percentage by which the CRE computed from the 1D-RT + true clouds or
312 1D-RT + retrieved clouds experiment exceeds the CRE computed from the 3D-RT + true clouds experiment
313 and imply that the 1D experiment overestimate the CRE relative to the 3D-RT + true clouds experiment.

314

315 **3. Results and discussion**

316

317 **3.1. Investigating the 3D radiative effects on simulated radiances**

318 Focusing first on SQ 1, we compare the reflectance obtained from the 1D and 3D RT simulations to
319 assess the impact of the 3D radiative effects on the reflectance radiation field, i.e., Box B vs. Box C in the
320 framework of **Fig. 1**. Specifically, we will investigate the differences in the reflectance simulated based on
321 3D and 1D RT (R^{3D-1D}) at the two λ (0.66 and 2.13 μm) required for our bi-spectral retrieval for both
322 low Sun (SZA 60°) and high Sun (SZA 5°) cases. To describe the 3D radiative effects on the observed
323 reflectance, classifications are made based on the increase in the brightness of a pixel in the LES domain.
324 A pixel in the LES domain is considered “illuminated” (“shadowed”) if its 3D RT-based reflectance is higher
325 (lower) than its 1D counterpart.

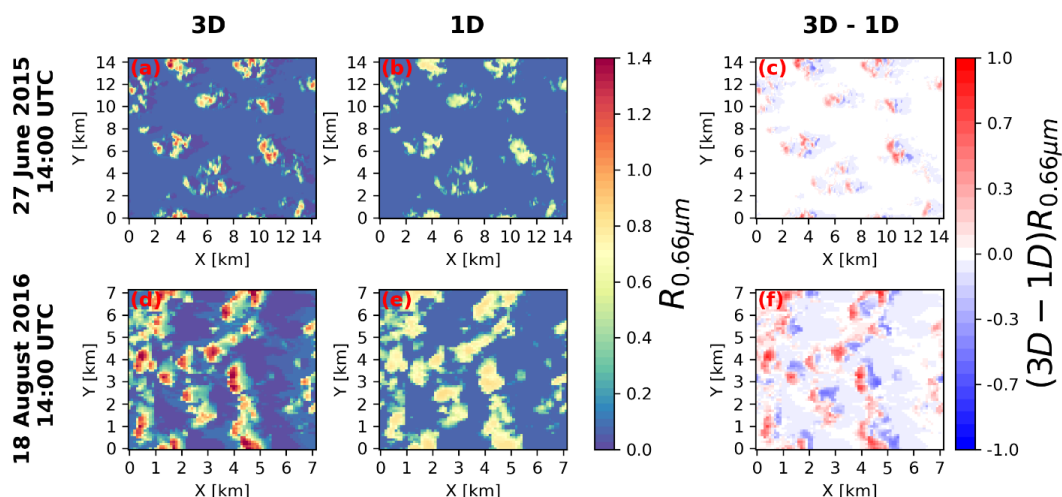
326 The 0.66 μm reflectance of the two cloud cases, based on 3D and 1D RT, and their differences are
327 shown in **Fig. 4** for the low sun condition and in **Fig. 5** for the high sun condition. In the low Sun case, the
328 3D RT-based simulated reflectance maps (**Fig. 4a** for the 27 June case and **Fig. 4d** for the 18 August case)
329 demonstrate the distinct presence of cloud shadows projected onto the clear-sky located on the opposite
330 cloud side and opposite to the Sun’s position located on the left of the domain. These features are absent
331 in the 1D RT-based simulated reflectance (**Fig. 4b, e**) because the radiation-cloud interactions in the 1D
332 reflectance field is dictated by the 1D RT theory, whereby the RT in each column is independent of the
333 surrounding column and thus the cloud shadow is cast directly beneath the cloud. Similar reflectance
334 characteristics are observed in the simulated 1D and 3D reflectance for the SWIR band (not shown). The
335 deviation of the 1D RT-based simulated reflectance from the 3D RT-based simulated reflectance leads to
336 R_{λ}^{3D-1D} with distinct pattern of illumination and shadowing observed in some pixels across the LES
337 domain for both the VNIR and SWIR bands ($R_{\lambda=0.66\mu\text{m}}^{3D-1D}$ shown in **Fig. 4c** and **f**). A closer examination of the
338 reflectance within cloudy regions in the low Sun case for the two cloud fields reveals a consistent pattern;
339 the illuminated pixels, where $R_{\lambda=0.66\mu\text{m}}^{3D-1D}$ is positive, are predominantly observed in sunlit regions that
340 directly face the Sun (e.g., X=3.5 km, Y=14 km in **Fig. 4c**). On the other hand, shadowed pixels, where
341 $R_{\lambda=0.66\mu\text{m}}^{3D-1D}$ is negative, are observed on the opposite side of the cloud layer (e.g., at X=5 km, Y=14 km in
342 **Fig. 4c**). These findings are consistent with previous 3D radiative effects studies for oblique solar geometry
343 [e.g., Várnai and Davies, 1999; Várnai, 2000; Marshak et al., 2006]. The observed opposing effects of
344 illuminating and shadowing in the low Sun angle case does not only depend on the orientation of the
345 cloud towards or away from the Sun, other factors like cloud-cloud interactions, cloud geometry and



346 aspect ratio, spatial distribution of the cloud in the domain and the horizontal transport of photons also
 347 contribute to these behaviors [Várnai and Marshak, 2001, 2002; Marshak and Davis, 2005; Marshak et al.,
 348 2006; Zhang et al., 2012].

349 In the case of the high Sun, the Sun is almost perpendicular (at SZA 5°), and its radiation interaction
 350 with clouds under 3D RT is different from that of the low Sun case. In 3D RT, when photons from the high
 351 Sun strikes a cloud, some photons are scattered and some leak from optically thick to optically thin cloudy
 352 regions and even out of the cloud layer [O’Hirok and Gauiter, 1998]. These leaking results in the darkening
 353 of the thick clouds and brightening of the surrounding thin clouds (Fig. 5a, d). Due to the absence of the
 354 photon leaking/darkening in the 1D RT, the 1D RT-based simulated reflectance (Fig. 5b, e) appear brighter
 355 than its 3D counterpart. Hence, results in negative $R_{\lambda=0.66\mu\text{m}}^{3\text{D}-1\text{D}}$ in the LES domain (Fig. 5c and f). The
 356 darkening characteristics is more pronounced in the 18 August case because it consists of a larger
 357 distribution of thicker clouds compared to the 27 June cloud field; large number of photons leaking from
 358 optically thicker clouds results in more significant reduction in the reflectance values and more prominent
 359 darkening effect than photons leaking from optically thinner clouds. Similar findings regarding the
 360 darkening characteristics are observed for the $2.13\mu\text{m}$ band, although these specific results are not
 361 depicted in the figures presented.

362



363

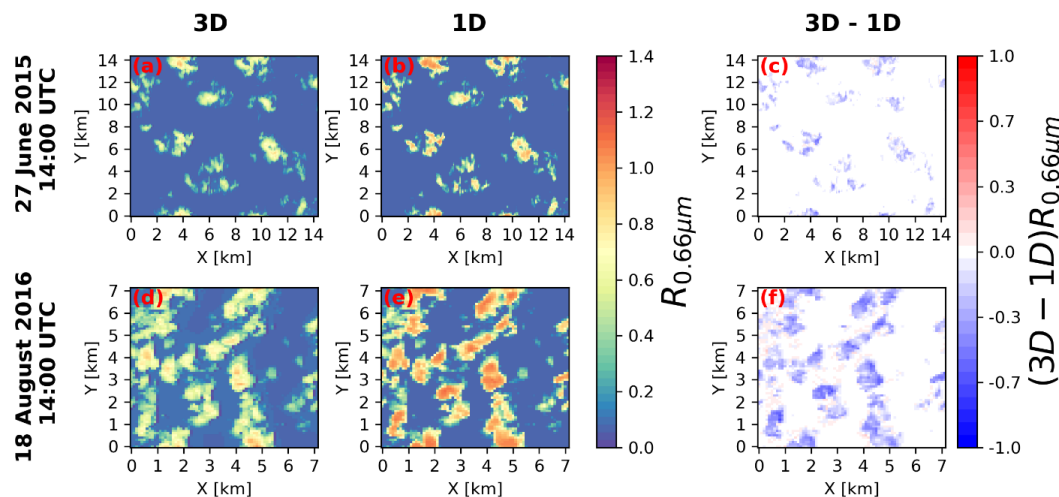
364 Fig. 4. Maps of the simulated 3D (a and d) and 1D (b and e) VNIR reflectance for the 27 June, and 18 August cases. Reflectance
 365 bias (3D-1D) are shown in (c) and (f) for the 27 June and the 18 August cases respectively. The direction of view is at nadir. The
 366 SZA is 60 degrees, and the Sun is at the left of the domain.

367



368

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370

371 **Fig. 5.** Maps of the simulated 3D (a and d) and 1D (b and e) VNIR reflectance for the high Sun case (SZA=5 degrees) for the 27
 372 June and 18 August 2016, 14:00 UTC cases. Reflectance bias (3D -1D) are shown in (c) and (f) for the 27 June and the 18 August
 373 cases respectively. The direction of view is at nadir. The Sun is almost perpendicular to the domain but slightly tilted to the left.

374 To examine the statistical characteristics of the reflectance bias in the LES domain, the probability
 375 density function (PDF) of the reflectance bias for “cloudy only” pixels are analyzed to investigate the 3D
 376 radiative effects on the observed cloud reflectance. Subsequently, we compared this PDF to that of the
 377 reflectance bias for both “cloudy and clear-sky” pixels (i.e., the whole LES domain) to highlight the effects
 378 of cloud presence on the overall reflectance bias within the LES domain.

379 The PDFs of R_{λ}^{3D-1D} for cloudy only pixels in the low Sun case (broken black and gray lines in **Fig.**
 380 **6a, b**) are characterized by positive and negative distribution in both VNIR and SWIR bands (corroborating
 381 the illuminating and shadowing effects in **Fig. 4c, f**). The overall positive reflectance bias observed in the
 382 VNIR and SWIR bands (domain mean reflectance bias of 0.0351 (0.0292) for the VNIR (SWIR) band in the
 383 27 June case and 0.0379 for the VNIR band in the 18 August case) indicates that the illumination effects is
 384 predominant when only cloudy pixels are considered. Meanwhile, a negative SWIR reflectance bias of
 385 -0.0233 is observed in the 18 August case. This negative reflectance bias is due to more photons being
 386 trapped in the 3D RT reflectance (likely due to its larger droplet sizes absorbing more photons) than in the
 387 1D RT which does not allow for horizontal photon transport. On the other hand, the PDFs of R_{λ}^{3D-1D}
 388 for the cloudy and clear-sky pixels (broken black and gray lines in **Fig. 6c, d**) is almost similar to that of the
 389 cloudy only but shows a shift of the distribution leftwards, almost centered around zero. This is expected
 390 because clear-sky regions not in the vicinity of any clouds exhibit negligible 3D radiative effects, which
 391 causes the distribution to shift closer to zero, since the cloud fraction for both cloud cases is less than 50
 392 %. The horizontal movement of photons from cloudy to surrounding clear-sky regions increase the 3D
 393 reflectance of clear-sky areas around the sunlit cloudy regions but the strong shadowing effects on the
 394 clear-sky region located opposite the sunlit direction dominates the clear-sky only areas, and results in a



395 negative mean bias when the reflectance of clear-sky only pixels are examined. Interestingly, the mean
396 reflectance bias for the cloudy and clear-sky pixels are of the same sign with the cloudy only values, these
397 indicates that the cloudy pixels have significant effect on the domain-scale statistics.

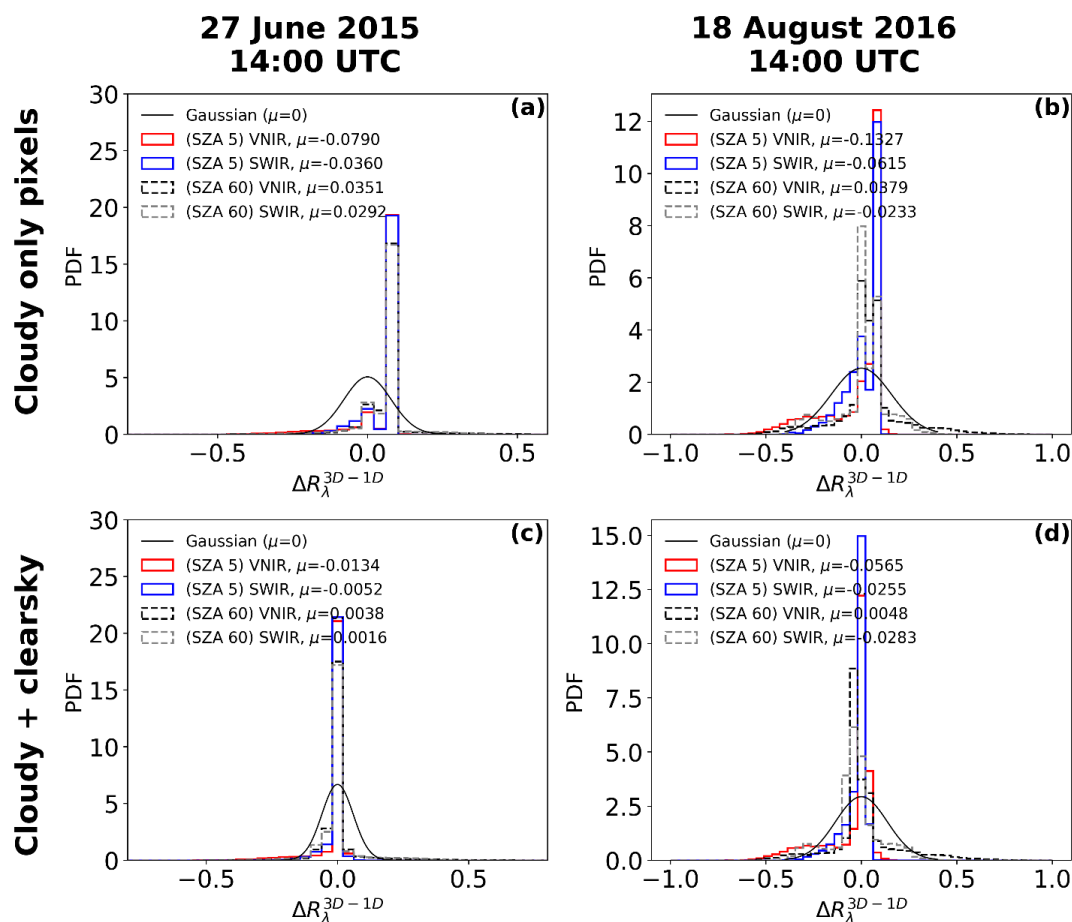
398 The PDFs of R_{λ}^{3D-1D} in the case of the high Sun for cloudy only pixels show a larger distribution of
399 pixels with positive R_{λ}^{3D-1D} in both the VNIR and SWIR band accompanied by longer tails to the left (red
400 and blue solid lines in **Fig. 6a, b**). However, the reflectance bias for both cloud cases present negative
401 values in the VNIR and SWIR bands. These observations suggest that large radiation/photos leak from a
402 small number of thick cloud pixels to a larger number of thin clouds. This phenomenon therefore
403 increases the number of thin clouds with positive reflectance bias, although of very small magnitude
404 when compared to the negative biases.

405 Similar to the low Sun case, the PDF of R_{λ}^{3D-1D} when both cloudy and clear-sky pixels for the high
406 Sun case are considered (red and blue solid lines in **Fig. 6c, d**), shows a significant distribution of values
407 close to zero. Due to the leaking of photons from thick clouds to thin clouds and clear-sky regions
408 surrounding the clouds, there is an increase in the 3D reflectance of clear-sky regions. Additionally, when
409 the Sun is high at SZA of 5° , there are very minimal shadows cast on the clear-sky regions. These two
410 highlighted reasons result in a positive reflectance bias for the clear-sky only region. Thus, the negative
411 value of the cloudy and clear-sky reflectance bias (same sign as the cloudy only) indicates that the domain
412 scale reflectance bias is dominated mainly by the cloudy only pixels and they play a significant role in the
413 domain-scale statistics.

414



415



416

417 **Fig. 6.** PDF (Probability density function) of reflectance bias ($\Delta R^{(3D-1D)}$) for cloudy only pixels for the 27 June case (a) and 18
 418 August case (b). PDF of reflectance bias for cloudy and clear-sky pixels for the 27 June case (c) and 18 August case (d). μ is the
 419 domain mean reflectance bias. A gaussian distribution (solid black curve) centered around zero is shown in all panels.

420

421 3.2. Investigating the 3D radiative effects on cloud retrievals

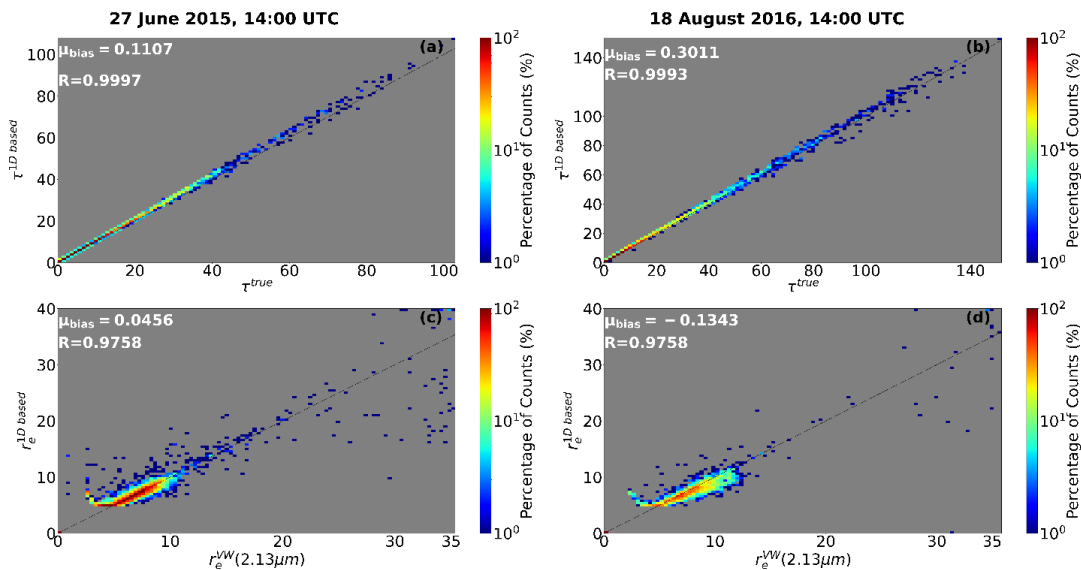
422 Focusing on SQ 2 in this section, we investigate how the reflectance bias, as discussed in the previous
 423 section, affect r_e and τ retrievals (i.e., Box A vs. Box D in the framework of **Fig. 1**). We utilize the 1D
 424 RT-based simulated reflectance as inputs for the LUT (explained in Sect. 2.3) to retrieve the 1D RT-based
 425 cloud droplet effective radius ($r_e^{1D \text{ based}}$) and cloud optical thickness ($\tau^{1D \text{ based}}$). Additionally, we use the
 426 3D RT-based simulated reflectance as inputs for the LUT to retrieve the 3D RT-based cloud droplet
 427 effective radius ($r_e^{3D \text{ based}}$) and cloud optical thickness ($\tau^{3D \text{ based}}$).



428 Before discussing analysis of the 3D and 1D RT-based retrievals comparison, we first check the accuracy
 429 of our retrievals by comparing the original LES cloud properties with our 1D RT-based retrievals (i.e.,
 430 comparing retrievals from 1D radiance in Box C with cloud properties in Box A in **Fig. 1**). For this purpose,
 431 the τ from the original LES (τ^{true}) is the vertical integration of the visible (0.66 μm) extinction coefficient
 432 of each column from cloud base to cloud top. For the LES r_e , we follow Zhang et al. [2017] analytical
 433 vertical weighting function (see their equation 4) to get the vertically weighted cloud droplet effective
 434 radius (r_e^{VW}) where the $\mu_0 = 0.5$, $\mu=1$ and the vertically weighting function parameter (b) associated with
 435 the 2.13 μm band was set to 2 to allow for a deeper penetration depth and for better correlation between
 436 the r_e^{VW} (2.13 μm) and bi-spectral retrievals.

437 **Fig. 7** shows the comparison between the r_e^{VW} (2.13 μm) and the $r_e^{\text{1D based}}$ as well as τ^{true} with the
 438 $\tau^{\text{1D based}}$ for the two cloud fields at SZA=60° and VZA=0°. For this comparison, the mean τ and r_e biases
 439 are $\mu_{\tau \text{ bias}} = \langle \tau^{\text{1D based}} - \tau^{\text{true}} \rangle$ and $\mu_{r_e \text{ bias}} = \langle r_e^{\text{1D based}} - r_e^{\text{VW}}(2.13 \mu\text{m}) \rangle$

440



441

442 **Fig. 7.** Joint histogram of bi-spectral retrieved τ based on 1D RT simulated reflectance $\tau^{\text{1D based}}$ vs Vertically integrated τ (τ^{true})
 443 for the 27 June case (a) 18 August case (b). Joint histogram of bi-spectral retrieved r_e based on 1D RT simulated total reflectance
 444 ($r_e^{\text{1D based}}$) vs vertically weighted cloud effective radius (r_e^{VW} (2.13 μm)) in (c) and (d). $\mu_{\text{bias}} = \langle \text{retrieved cloud property} -$
 445 $\text{reference cloud property} \rangle$.

446 For the two cloud fields considered in this study, the $\tau^{\text{1D based}}$ is highly correlated with the τ^{true}
 447 as seen in the joint histogram plots (**Fig. 7a** and **b**) with a correlation coefficient (R) of 0.9997 for the 27
 448 June case, and a R of 0.9993 for the 18 August case, although both have a slight positive mean bias
 449 ($\mu_{\tau \text{ bias}} = 0.1107$ and 0.3011 for the 27 June and 18 August cases respectively). Also, the comparisons of
 450 the $r_e^{\text{1D based}}$ with the r_e^{VW} (2.13 μm) in **Fig. 7c** and **d**, shows good correlation ($R > 0.96$) for both cloud
 451 cases, and slightly positive mean biases ($\mu_{r_e \text{ bias}} = 0.0456$) for the 27 June case and a negative mean bias
 452 ($\mu_{r_e \text{ bias}} = -0.1343$) for the 18 August case. Certain extreme outliers bias is observed in the r_e
 453 comparisons, these outliers are attributed to thin clouds and have been studied by Miller et., al [2017].



454 Several studies [e.g., Miller et al., 2016, 2017; Zhang et al., 2012] have investigated the accuracy of 1D
455 bi-spectral retrievals compared to vertically weighted retrievals as well as the impact of cloud vertical
456 profile on bi-spectral retrievals. Since we have good agreement between retrievals from the 1D RT-based
457 reflectance and the original LES cloud field properties, this study will use the $r_e^{1D\text{ based}}$ and $\tau^{1D\text{ based}}$ as
458 the reference cloud properties and directly compare them with the $r_e^{3D\text{ based}}$ and $\tau^{3D\text{ based}}$ to investigate
459 the impacts of the 3D radiative effects on the retrievals.

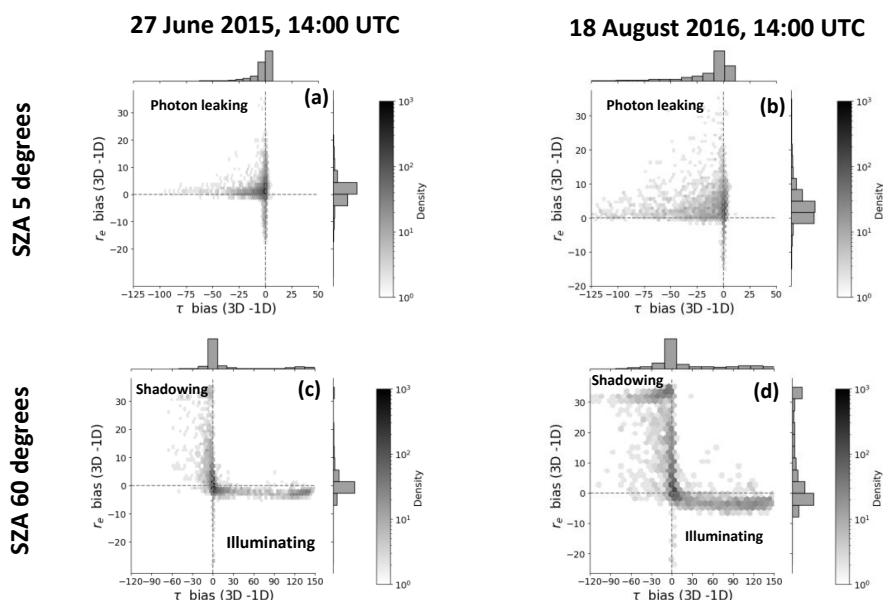
460 For the high Sun case, under 3D RT, photons leaking from optically thick regions to optically thin
461 cloudy regions make the 3D radiance field to appear darker than its 1D counterpart. Consequently, for
462 retrievals, the reflectance observation on the LUT space shifts leftwards and downwards. Thereby, larger
463 $r_e^{3D\text{ based}}$ are retrieved and the retrieved $\tau^{3D\text{ based}}$ are smaller. These overestimation of the $r_e^{3D\text{ based}}$ and
464 underestimation of the $\tau^{3D\text{ based}}$ retrievals is evident in the joint histogram plots of the r_e biases
465 ($r_e^{3D\text{ based}} - r_e^{1D\text{ based}}$) against the τ biases ($\tau^{3D\text{ based}} - \tau^{1D\text{ based}}$) for the two cloud fields considered in this
466 study. In **Fig. 8a** and **b** for the 27 June and 18 August cases respectively, we observe that a larger portion
467 of the retrievals bias falls in the third quadrant (anticlockwise starting from bottom right quadrant) where
468 the τ bias is negative (underestimation of $\tau^{3D\text{ based}}$) and r_e bias is positive (overestimation of $r_e^{3D\text{ based}}$).

469 For the retrievals from the low Sun case, under 3D RT, it is generally observed that the sunlit
470 optically thick cloud regions experiences illuminating effects. In the LUT solution space, illumination
471 phenomena will shift the observed reflectance upwards and rightwards where the LUT r_e grid isolines
472 represents smaller droplets sizes and the LUT τ isolines represents thicker clouds. Thus, $\tau^{3D\text{ based}}$ are
473 larger and $r_e^{3D\text{ based}}$ are smaller than their 1D counterpart (as seen in the first bottom right quadrant in
474 **Fig. 8c** and **d** for the 27 June and 18 August cases respectively). If a pixel is shadowed, the reflectance
475 observation on the LUT solution space will shift downwards and leftwards to regions where the LUT r_e
476 grid isolines represent larger droplet sizes and the LUT τ isolines represents thinner clouds. These leads
477 to larger $r_e^{3D\text{ based}}$ while the $\tau^{3D\text{ based}}$ are underestimated compared to its 1D counterpart (as seen in the
478 third quadrant anticlockwise from the bottom right in **Fig. 8c** and **d** for the 27 June and 18 August cases
479 respectively). τ and r_e retrieval biases in satellite observations have been investigated in numerous
480 studies [e.g., Marshak et al., 1995; Várnai and Marshak, 2002; Zhang and Platnick, 2011; Zhang et al.,
481 2011, 2012], and in common occurrence, overestimation of the retrieved τ is coupled with the
482 underestimation of the retrieved r_e and vice versa.



483

484



485

486 **Fig. 8.** Bias in cloud effective radius (r_e) against bias in cloud optical thickness (τ) retrievals for Solar zenith angle (SZA) 5 degrees
 487 for (a) the 27 June and (b) 18 August case. Bias in r_e against bias in τ retrievals for SZA 60 degrees for (c) the 27 June and (d) 18
 488 August case.

489

490 **Table 3** shows the frequency of failed and successful retrievals from 3D RT-based simulated
 491 reflectance observation for the two cloud fields considered in this study. It is observed that the number
 492 of failed retrievals is small for the SZA 5° case, while the retrieval failures are larger for the SZA 60° case
 493 for both cloud fields under consideration. This is due to a significant number (> 40 %) of the VNIR-SWIR
 494 reflectance observations falling outside the LUT solution space when the Sun is low. Conversely, the small
 495 retrieval failure rates (< 13 %) for the high Sun case is because most of the VNIR-SWIR reflectance
 496 observation (> 85 %) falls within the LUT solution space.

497

498 Table 3. Statistics of successful and failed retrievals from the 3D based radiance for the 27 June and 18 August cloud fields at
 499 Solar zenith angle (SZA) 5 and 60 degrees. The columns from left to right are Case name (Identified by date and time), solar
 500 zenith angle (SZA), Number of pixels with successful retrievals only, Pixels with failed retrievals, Total number of successful and
 501 failed retrievals.

Case name	SZA	No of pixels with successful retrievals only	Pixels with failed retrievals		Total number of successful and failed pixels
			Category of failed retrievals	No of pixels	
27 June 2015, 14:00 UTC	5°	3670 (87.82 %)	r_e too large	85 (2.03 %)	509 (12.17%)
			r_e too small	365 (8.73 %)	
			τ failures	63 (1.41 %)	
	60°	2100 (50.16%)	r_e too large	97 (2.32 %)	2079 (49.74%)
r_e too small	1035 (24.77 %)				



			τ failures	947 (22.66 %)		
18 August 2016, 14:00 UTC	5°	2344 (96.02%)	r_e too large	46 (1.88 %)	97 (3.97%)	2441 (100 %)
			r_e too small	29 (1.188 %)		
			τ failures	22 (0.90 %)		
	60°	1368 (56.04%)	r_e too large	339 (13.88 %)	1073 (43.96%)	2441 (100 %)
			r_e too small	178 (7.29 %)		
			τ failures	556 (22.77 %)		

Values in parentheses are percentage of counts. (Percentage of counts = Number of affected pixels/ Total number of pixels)

502

503 3.3. Investigating the 3D radiative effects on the Broadband radiative flux

504

505 3.3.1. Investigating the 3D radiative effects on the broadband radiative flux: Using a 506 combination of the successful and failed retrievals as the input cloud property

507 Focusing on SQ3 in this section, we will compare the broadband SW flux results from the 1D-RT +
508 retrieved clouds experiment with those from the 3D-RT + true clouds experiments (i.e., Box E vs. Box F in
509 **Fig. 1**) to investigate the impact of cloud retrieval biases due to the 3D radiative effects on the broadband
510 SW radiative flux. We will also compare results from the 1D-RT + true clouds experiment to those from
511 the 3D-RT + true clouds experiment (i.e., Box F vs. Box G in **Fig. 1**) to study the impact of neglecting
512 horizontal photons transport on the broadband SW flux results.

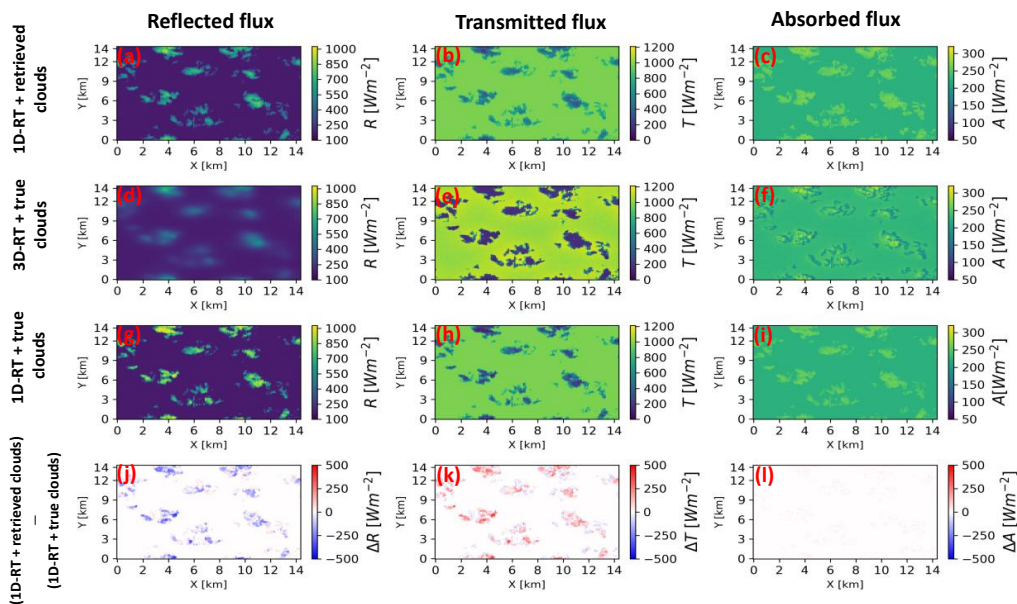
513 It is important to note here that both the successful and the failed retrievals as described in Sect.
514 2.4 are included in the RT simulations in the control simulations presented in this section. The motivation
515 for including the failed retrievals is to preserve the impacts of this significant fraction of pixels on the
516 domain averaged fluxes and CRE simulations, even though the retrieval of τ and r_e based on the bi-
517 spectral method fails for them. In addition to the controlled simulations, we have also conducted
518 sensitivity studies, where we exclude the failed retrievals in the analysis. The results are shown and
519 discussed in the Appendix.

520 Maps of the simulated SW broadband radiative quantities (reflected flux at the TOD, transmitted
521 flux at the surface, and column absorbed flux) for the 27 June case at the high and low Sun angles are
522 presented in **Fig. 9** and **Fig. 10**, respectively. These figures reveal several interesting and important points.
523 First, it is interesting to note that the reflected flux in **Fig. 9d** seems blurry in comparison with 1D results
524 in **Fig. 9a** and **Fig. 9g**. The same is also seen in **Fig. 10a** and **g**. This is because in 1D RT, simulation of the
525 upwelling hemispheric flux at a given point at the TOD is determined *only* by the cloud and surface
526 properties in the column beneath such point. In contrast, in 3D RT simulation, it depends on the cloud and
527 surface properties of both the corresponding column and a large extent of the surrounding columns, as a
528 result of simple parallax effect. Therefore, the contrast between two adjacent columns in the 1D
529 simulation, for example, a cloudy column and an adjacent clear-sky column next to it, is quite large,
530 whereas the contrast for the same two columns in 3D simulation is much smaller because the two have a
531 significant overlap in terms of the areas that have influences on their flux. Because of this fundamental
532 difference between 1D and 3D simulations, a pixel-to-pixel comparison of the upwelling flux is not
533 appropriate. Instead, we compare the domain-averaged statistics. Before we delve into that, we first aim
534 to unravel how cloud property retrieval errors affect 1D RT flux solutions. For this purpose, we compare
535 the TOD reflected flux as well as the transmitted flux at the surface from the 1D-RT + retrieved clouds and



536 1D-RT + true clouds experiments (since they are both computed via 1D RT). The TOD reflected flux from
 537 the 1D-RT + retrieved clouds experiments have visible signatures of the input cloud property retrievals.
 538 For instance, in the high Sun case, smaller reflected flux values (recall, underestimated τ dominates
 539 retrievals from high Sun radiance) dominate the reflected flux results from the 1D-RT + retrieved clouds
 540 experiment (**Fig. 9a**) as compared to the reflected flux from the 1D-RT + true clouds experiment (**Fig. 9g**).
 541 The underestimation of the reflected flux in the 1D-RT + retrieved clouds experiment compared to the
 542 1D-RT + true clouds experiment is evident in **Fig. 9j**. This difference is also well captured in the
 543 domain-averaged values which will be discussed later in this section. In the low Sun case, comparison
 544 between the reflected flux from the 1D-RT + retrieved clouds experiment and corresponding flux from the
 545 1D-RT + true clouds experiment reveals that in the 1D-RT + retrieved clouds, the overestimated retrieved
 546 τ areas characterized by thicker clouds (i.e., retrieved from illuminated pixels) provides larger reflected
 547 flux values and the underestimated retrieved τ areas characterized by thinner clouds (i.e., retrieved from
 548 shadowed pixels) have smaller reflected flux values than their 1D-RT + true clouds counterpart (**Fig. 10j**).
 549 Their overall effect on the domain reflected flux values depends on how the opposite 3D radiative effects
 550 (cloud side illumination and shadowing) mitigate each other.

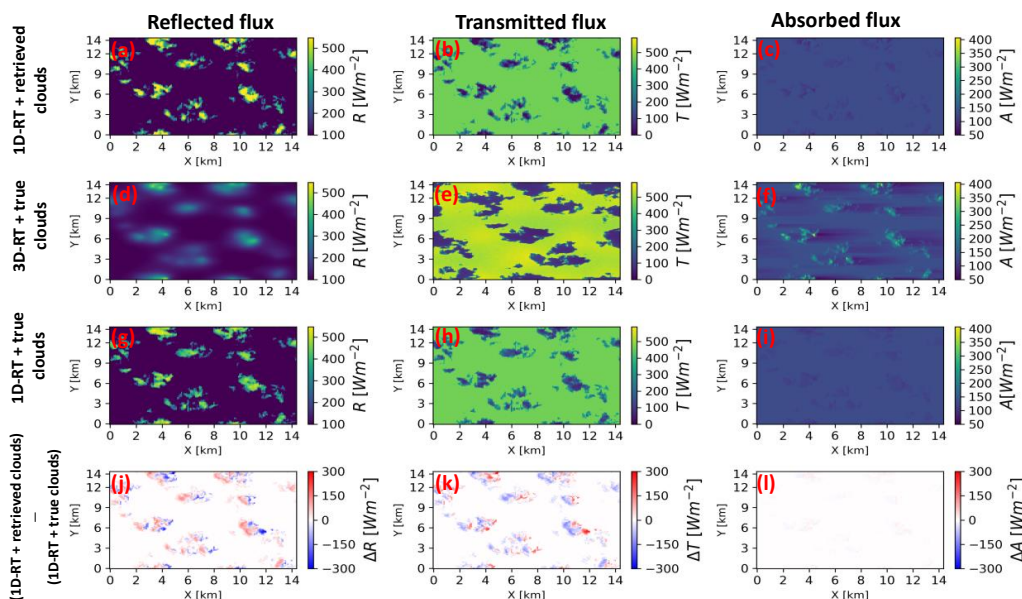
551 An examination of the transmitted flux at the surface between the 1D-RT + true clouds and 1D-RT +
 552 retrieved clouds experiments for the high Sun case reveals that the transmitted flux at the surface beneath
 553 clouds in the 1D-RT + retrieved clouds experiment is larger compared to results from the 1D-RT + true
 554 clouds experiment, while they have same values in clear-sky regions (**Fig. 9k**). This is expected since the
 555 reflected flux from the 1D-RT + retrieved clouds experiment is lesser than results from the 1D-RT + true
 556 clouds experiment. Thus, the amount of flux at the surface beneath the cloud in the 1D-RT + retrieved
 557 clouds experiment increases. For the low Sun case, the transmitted flux at the surface beneath the clouds
 558 in the 1D-RT + true clouds and 1D-RT + retrieved clouds experiments have higher values where the TOD
 559 reflected flux is low and lower values where the TOD reflected flux is high (**Fig. 10k**).



560



561 **Fig. 9.** Simulated shortwave broadband Reflected, Transmitted and Absorbed fluxes for 1D-RT + retrieved clouds (a)-(c), 3D-RT +
 562 true clouds (d)-(f), 1D-RT + true clouds (g)-(i) and difference between 1D-RT + retrieved clouds and 1D-RT + true clouds (j)-(l) for
 563 Solar zenith angle=5° and View zenith angle=0°. Sun is high and slightly on the Left-hand side of the domain. The solar irradiance
 564 at the top of the domain (TOD) scales with the cosine of the Solar zenith angle.



565

566 **Fig. 10.** Simulated shortwave broadband Reflected, Transmitted and Absorbed fluxes for 1D-RT + retrieved clouds (a)-(c), 3D-RT
 567 + true clouds (d)-(f), 1D-RT + true clouds (g)-(i) and difference between 1D-RT + retrieved clouds and 1D-RT + true clouds (j)-(l)
 568 for Solar zenith angle=60° and View zenith angle=0°, Sun is on the Left-hand side of the domain. The solar irradiance at the top
 569 of the domain (TOD) scales with the cosine of the Solar zenith angle.

570 The radiative quantities across the LES domain have different characteristics in the 3D-RT + true
 571 clouds experiment stemming from the horizontal transport of photons across pixels. In the high Sun case
 572 of the 3D-RT + true clouds experiment, due to the leaking of photons through cloud sides, the diffused
 573 transmitted flux at the surface is larger than results obtained from the 1D-RT + true clouds experiment,
 574 similar to the findings made by Wapler and Mayer, [2008]. This is primarily why the transmitted flux at
 575 the surface in the high Sun case of the 3D-RT + true clouds experiment is higher than corresponding 1D-
 576 RT + true clouds and 1D-RT + retrieved clouds values, especially around cloud edges (**Fig. 9b, e, h**). For
 577 the low Sun case of the 3D-RT + true clouds experiment (**Fig. 10e**), due to 3D RT, there is an increase in
 578 the total effective cloud cover [Di Giuseppe and Tompkins, 2003; Tompkins and Di Giuseppe, 2007], as
 579 well as an increase in the size of the cloud shadow, which reduces the transmitted flux at the surface (e.g.,
 580 around $[X=7, Y=6 \text{ km}]$ in **Fig. 10e**). Meanwhile, the cloud shadowed areas projected on the surface shift
 581 according to the SZA [Walper and Mayer, 2008; Wissmeier et al., 2013; Jakub and Mayer, 2015, 2017] and
 582 are larger for more oblique SZAs.

583 Table 4. . Domain-averaged broadband shortwave (SW) reflected flux at the top of the domain (R), transmitted flux at the
 584 surface (T) and absorbed flux (A) from the 1D-RT + retrieved clouds, 1D-RT + true clouds and 3D-RT + true clouds experiments
 585 for the two cloud cases at solar zenith angle (SZA) 5 degrees and SZA 60 degrees.

	SZA 5 degrees	SZA 60 degrees
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Case Name		1D-RT + retrieved clouds (Wm^{-2})	+ 3D-RT + true clouds (Wm^{-2})	1D-RT + true clouds (Wm^{-2})	1D-RT + retrieved clouds (Wm^{-2})	+ 3D-RT + true clouds (Wm^{-2})	1D-RT + true clouds (Wm^{-2})
27 June 2015 (14:00 UTC)	R	215.44 (213.94)	215.93	225.37 (223.52)	134.22 (111.21)	137.87	133.04 (112.01)
	T	918.97 (920.68)	918.79	910.76 (912.88)	419.60 (441.77)	414.36	420.97 (441.34)
	A	228.56 (228.37)	228.23	226.82 (226.60)	130.25 (131.13)	131.82	130.11 (130.79)
18 Aug. 2016 (14:00 UTC)	R	315.16 (316.82)	308.68	355.26 (357.12)	209.74 (174.40)	218.62	211.54 (171.59)
	T	805.34 (803.59)	812.25	770.21 (768.26)	342.50 (378.46)	326.53	341.92 (382.68)
	A	242.36 (242.48)	241.95	237.36 (237.46)	131.74 (131.20)	138.86	130.55 (129.76)

Note: Values before the parentheses are calculated from the combination of failed and successful retrievals representing the total cloudy population, while values in parentheses are calculated from successful retrievals only representing the total cloudy population. clear-sky pixels values have been included in all calculations.

586

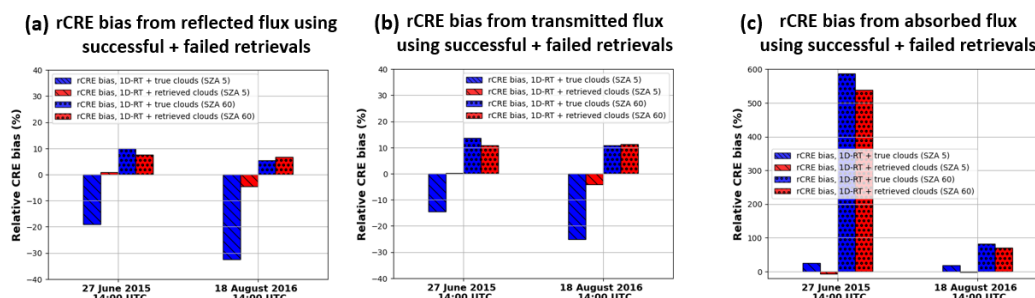
587 The domain-averaged broadband reflected flux at the TOD, transmitted flux at the surface, and
 588 column absorbed flux values from the 1D-RT + retrieved clouds, 1D-RT + true clouds and 3D-RT + true
 589 clouds experiments for the 27 June and 18 August cases at SZA 5° and SZA 60° are reported in **Table 4**. As
 590 previously explained, the predominant photon leaking associated with high Sun 3D RT and the ensuing
 591 underestimation of the retrieved τ which dominate the cloud property retrievals from high Sun 3D
 592 simulated reflectance, increases the number of retrieved optically thinner clouds (relative to the original
 593 LES τ used in the 1D-RT + true clouds calculations) utilized as inputs for the 1D-RT + retrieved clouds
 594 calculations. This leads to the underestimation of the domain-averaged 1D-RT + retrieved clouds reflected
 595 flux compared to the 1D-RT + true clouds reflected flux; In the 27 June case, the domain-averaged
 596 reflected flux from the 1D-RT + retrieved clouds experiment ($215.44 Wm^{-2}$) is underestimated compared
 597 to the corresponding 1D-RT + true clouds value ($225.37 Wm^{-2}$) by about $9.93 Wm^{-2}$. While in the 18
 598 August case, the 1D-RT + retrieved clouds experiment underestimates the domain-averaged reflected flux
 599 ($315.16 Wm^{-2}$) compared to the corresponding 1D-RT + true clouds value ($355.26 Wm^{-2}$) by 40.1
 600 Wm^{-2} . The larger value of the underestimated domain-averaged reflected flux in the 18 August case
 601 stems from its larger cloud fraction and τ bias. The transmitted flux at the surface below clouds is
 602 dependent on the amount of flux reflected towards the TOD; lower reflected flux values indicate that less
 603 radiation is reflected from the clouds, which allows for a greater amount of radiative flux to be transmitted
 604 to the surface beneath the clouds. This reason, coupled with the overestimation of the transmitted flux
 605 at the surface due to missed thin clouds in our bi-spectral retrievals (red regions in **Fig. 3**, retrieved $\tau = 0$
 606 for VNIR reflectance less than the smallest LUT τ value), explains why for the high Sun case, the
 607 domain-averaged values of the surface transmitted flux are higher in the 1D-RT + retrieved clouds
 608 experiment compared to the 1D-RT + true clouds values, resulting in differences of $8.21 Wm^{-2}$ and 35.13
 609 Wm^{-2} for the 27 June and 18 August cases respectively. Although, for the high Sun angle, the
 610 contribution of the missed thin clouds to the overestimation of the transmitted flux at the surface beneath
 611 cloud in our case study is small (Constituting about 0.23% and 0.34% of the domain-averaged surface
 612 transmitted flux for the 27 June and 18 August high cases respectively).

613 Comparing results from the three sets of experiments in **Table 4** reveals that for the high Sun case,
 614 the results from the 1D-RT + retrieved clouds clearly agree better with the benchmark 3D-RT + true clouds
 615 experiments, than the 1D-RT + true clouds. In the 27 June case, the domain-averaged biases in TOD
 616 reflected, surface transmitted and absorbed fluxes for the 1D-RT + retrieved clouds experiment are 0.49,
 617 -0.18 and $-0.33 Wm^{-2}$ respectively, which are significantly smaller in magnitude than those for the 1D-
 618 RT + true clouds experiment of -9.44 , 8.03 , and $1.41 Wm^{-2}$ respectively. Similarly, for the 18 August case,



619 the domain-averaged biases in reflected, transmitted and absorbed fluxes for 1D-RT + retrieved clouds
620 experiment are -6.48 , 6.81 and -0.41 Wm^{-2} respectively compared to corresponding biases of -46.58 ,
621 42.04 , and 4.59 Wm^{-2} for the 1D-RT + true clouds experiments. These results suggests that the 1D-RT +
622 retrieved clouds experiment gives an overall better radiative energy estimate than the 1D-RT + true clouds
623 experiment for the high SZA case. In the low Sun case, the two 1D RT experiments are very close to each
624 other and there is not a clear winner when compared to the benchmark 3D RT results. In the 27 June case,
625 the 1D-RT + retrieved clouds experiment agrees slightly better with 3D results than the 1D-RT + true clouds
626 experiment, but the opposite is true in the 18 August case. This result seems to suggest that although in
627 the low Sun case the illuminating and shadowing effects can lead to large retrieval biases, they tend to
628 cancel out each other in the flux computations. Interesting, both 1D results tend to underestimate the
629 reflected flux and overestimate the transmitted flux. This is probably because the illuminating effect is
630 dominant in the 3D RT leading to some extremely bright pixels. But they are not captured in the 1D
631 experiments, even in the 1D-RT + retrieved cloud experiment using the upper limit of $\tau=158.48$ in the flux
632 computation. Thus, the reflected flux quickly reaches the asymptotic value when τ is large and therefore
633 simply using larger τ value in 1D-RT cannot simulate the extreme brightness of clouds due the illuminating
634 effect in 3D RT.

635 Because both cloud cases have a cloud fraction lower than 50%, the domain-averaged statistics
636 include a large fraction of clear-sky pixels. Now we focus our scope on cloudy pixels and investigate the
637 differences in CRE. The rCRE bias provides a quantitative estimate of how these biases affect the CRE. For
638 the two cloud cases considered in this study, plots of rCRE bias computed from the reflected flux at the
639 TOD, transmitted flux at the surface and the absorbed flux at SZA 5° and 60° for the 1D-RT + true clouds
640 and 1D-RT + retrieved clouds experiments relative to the 3D-RT + true clouds experiments are presented
641 in **Fig. 11**. In the 27 June case, the rCRE bias of 0.97% computed from the reflected high Sun 1D-RT +
642 retrieved clouds results indicate a negligible deviation (less than 1 %) from the 3D-RT + true clouds CRE
643 while the rCRE bias of -19% computed from the reflected high Sun 1D-RT + true clouds result show that
644 the bias is quite substantial. Similarly, for the 18 August case, the rCRE bias computed from the reflected
645 high Sun 1D-RT + retrieved clouds experiment results is less than 5%. On the other hand, the rCRE bias of
646 -32.48% computed from the reflected high Sun 1D-RT + true clouds result show that the bias is quite
647 large. Similar results are obtained for rCRE bias computed from the surface transmitted flux. In the 27
648 June case, the rCRE bias computed from the 1D-RT + retrieved clouds transmitted flux is 0.33% (**Fig. 11b**,
649 second bar on the left) which shows minimal bias less than 1%, while the rCRE computed from the 1D-RT
650 + true clouds transmitted flux is -14.5% (**Fig. 11b**, first bar on the left). Similarly, for the 18 August case,
651 the rCRE bias computed from the transmitted 1D-RT + retrieved clouds result is -4.12% (**Fig. 11b**, second
652 bar on the right), while the rCRE bias computed from the transmitted 1D-RT + true clouds result is
653 -25.10% (**Fig. 11b**, first bar on the right).



654

655 **Fig. 11.** Relative cloud radiative effect (rCRE) bias computed from the successful + failed retrievals (a) top of the domain
656 reflected, (b) surface transmitted and (c) absorbed flux for the two cloud fields.

657 When the absorbed flux is taken into consideration, for the 27 June high Sun case, the rCRE bias computed
658 from the absorbed 1D-RT + retrieved clouds flux is -6.05% (**Fig. 11c**, second bar on the left) which is less
659 bias compared to the rCRE bias computed from the absorbed 1D-RT + true clouds flux is 25.64% (**Fig. 11c**,
660 first bar on the left). Similarly, for the 18 August case, the rCRE bias computed from the absorbed 1D-RT
661 + retrieved clouds flux is -1.73% (**Fig. 11c**, second bar on the right), while the rCRE bias computed from
662 the absorbed 1D-RT + true clouds flux is 19.09% (**Fig. 11c**, first bar on the right). For the low Sun case, the
663 rCRE biases from the two 1D-RT experiments are comparable, which is consistent with the domain-
664 averaged statistics in **Table 4**. Evidently, both 1D-RT experiments overestimate the CRE at TOD and
665 surface, which means an underestimation of cloud reflection and overestimation of transmission. This is
666 consistent with the results in **Table 4**.

667 Overall, the above analysis indicates that the 1D-RT + retrieved clouds experiment provides a better (in
668 the high Sun case) or at least comparable (in the low sun case) results than 1D-RT + true cloud experiment
669 for both domain-averaged flux statistics and CRE when compared to the benchmark 3D-RT results. With
670 these results we can conclude that despite the potential biases due to 3-D effects the retrieved cloud
671 properties based on 1-D RT from the bi-spectral method still provide a reasonable observational basis to
672 estimate the broadband flux and CRE.

673 4. Summary and Conclusion

674 It is well known that the bi-spectral cloud property retrievals based on the 1D-RT have significant
675 errors due to the 3D radiative effects. In this study, we investigate whether the biased retrievals can still
676 be used to estimate the broadband flux and CRE. To address this question, we selected two cloud fields
677 from the LASSO activity: one on 27 June 2015 and another on 18 August 2016 to serve as case studies for
678 our research. The LES cloud fields have different microphysics with different CBH, CTH and the value of
679 the cloud fraction for the 18 August 2016 cloud field (47.08%) is more than twice that of the 27 June 2015
680 (20.15%) cloud field. Radiance simulations, bi-spectral retrievals, and broadband SW flux radiative transfer
681 simulations were performed using these cloud fields at two SZAs, a high Sun case ($SZA=5^\circ$) and a low Sun
682 case ($SZA=60^\circ$) and the results were analyzed. The flux computations were carried out in three sets, the
683 reference broadband SW flux calculations were performed using the cloud properties from the original
684 LES cloud field under 3D RT (we call this the 3D-RT + true clouds experiment), we also computed similar
685 RT broadband SW flux calculations with the same cloud properties from the original LES cloud field except
686 that the RT calculations were computed using 1D RT (we call this the 1D-RT + true clouds experiment).



687 Additionally, we computed the last set of broadband SW flux calculations using 1D RT and bi-spectrally
688 retrieved cloud properties as inputs (we call this the 1D-RT + retrieved clouds experiment).

689 The high Sun radiance results, for the two cloud fields show that in 3D RT high Sun case, the photon
690 leaking from optically thick cloudy regions to optically thin cloudy regions dominate the LES reflectance
691 field. These results in overestimated r_e and underestimated τ dominating the cloud property retrievals.
692 While results from the low Sun case, for the two cloud fields considered show that in comparison to the
693 1D RT radiance fields, illuminating and shadowing effects both occur in the 3D RT simulated radiance
694 observation. Therefore, retrievals from the low Sun 3D radiance observations are characterized mainly by
695 both overestimation of τ and underestimation of r_e in illuminated pixels and underestimation of τ and
696 overestimation of r_e in shadowed pixels. The cumulative effects of these illuminating and
697 shadowing/Photon leaking effects and its impacts on the retrieved cloud properties dictates their impact
698 on the broadband radiative flux.

699 The results from the broadband SW radiative fluxes computation showed that although, the
700 bi-spectrally retrieved cloud properties are often biased due to the 3D radiative transfer effects, for high
701 Sun cases, calculations of the CRE from these 1D-RT + retrieved clouds values agree well with the
702 benchmark values (which is the 3D-RT + true clouds experiment in our case) with agreement within 7s%
703 for rCRE bias calculations from the reflected, transmitted and absorbed fluxes in the high Sun cases.
704 Conversely, the rCRE bias computed from the 1D-RT + true clouds flux quantities could reach about 33%.
705 Thus, for high Sun situations, the 1D-RT + retrieved clouds experiment provides consistently better
706 estimates of the CRE than the 1D-RT + true clouds experiment. For the low Sun case, the two 1D RT
707 experiments provide comparable results, both underestimating cloud reflection and overestimating
708 transmission, and there is not a clear winner when compared to the 3D RT benchmark.

709 The influence of the failed retrievals on the CRE was also investigated (see details in Appendix),
710 with results indicating that for the high Sun case, the impact of the failed retrievals on the radiative flux
711 quantities is negligible, with less than 6% changes observed in the rCRE bias computed from the
712 domain-averaged TOD reflected, surface transmitted and absorbed flux 1D-RT + retrieved clouds and 1D-
713 RT + true clouds experiments. Such is not the case for the low Sun case where the failed retrievals have a
714 very huge impact on the radiative flux quantities. Excluding the failed retrievals from the domain-averaged
715 reflected, transmitted, and absorbed flux 1D-RT + retrieved clouds and 1D-RT + true clouds low Sun case
716 analysis could increase the rCRE bias by a as much as factor of 6 compared to values which included the
717 failed retrievals in the analysis.

718 In conclusion, despite the potential biases due to the 3D radiative effects, the retrieved cloud
719 properties based on 1D RT from the bi-spectral method still provide a reasonable observational basis to
720 estimate the broadband flux and CRE. Some future questions that warrant answers involves how the 3D
721 radiative effects affect the broadband fluxes for different cloud arrangements and other types of clouds,
722 such as deep convective clouds. Also, while we have considered only nadir view angle in this work,
723 previous studies [e.g., Várnai and Marshak, 2007] have shown that the biases of 1D cloud retrievals vary
724 systematically with view direction, therefore, the impacts of off-nadir view directions on the broadband
725 flux need to be investigated. Another important studies will be to determine how changes in surface
726 albedo and type affects our results. Additionally, while our case study mainly focused on the impact of
727 the 3D radiative effects on SW fluxes, the impact of the 3D radiative effects on LW radiation is important
728 and need to be investigated.



729 Appendix A: Impacts of failed retrievals on the radiative flux

730 The 1D-RT + retrieved clouds experiment and domain radiative flux analysis in Sect. 3.3 utilized
731 both the successful and failed retrievals (categorized in Sect. 2.4) to represent the total population of
732 cloudy pixels. Henceforth, both successful and failed retrievals as a representative of the total population
733 of cloudy pixels will be referred to as “all retrieved cloud pixels”. In this appendix, our focus is to examine
734 and compare the TOD reflected, surface transmitted and column absorbed radiative fluxes, when the
735 failed retrievals are excluded from the radiative flux analysis. This will help to diagnose if using solely
736 successful retrievals as a representative of the total population of cloudy pixels in the LES domain will
737 produce the correct radiative energy estimates and thus provide information on the radiative properties
738 of the excluded failed retrievals.

739 An examination of the high Sun domain-averaged TOD reflected, surface transmitted and column
740 absorbed fluxes for both LES cloud cases, when only successful retrievals represent the total population
741 of cloudy pixels in the 1D-RT + retrieved cloud experiment, show minimal changes (within the range ± 1.9
742 Wm^{-2}) from previous values which utilized all retrieved cloud pixels in the radiative flux analysis (**Table**
743 **4**). This is due to the small number of failed retrievals in the high Sun scenario ($< 14\%$ for both cloud cases;
744 **Table 3**). But this is not the case for the low Sun case, where changes between the two aforementioned
745 calculations are large, reaching up to $\pm 35.96 \text{Wm}^{-2}$ (**Table 4**). These large changes are because of the
746 large number of failed retrievals from strong 3D radiative effects ($> 43\%$ for both cloud cases; **Table 3**) as
747 well as different radiative behavior of the failed retrievals categories observed in the low Sun scenario.
748 **Fig. A1** shows plots of successful and failed retrievals categories (classified as described in Sect. 2.4) from
749 the high and low Sun radiance for the 27 June and 18 August cases. From these plots, it is observed that
750 when the SZA is 60° , the r_e too small failures are predominant around cloud edges in the sunlit areas. The
751 τ failures are observed mostly in the illuminated sunlit cloudy regions and the r_e too large failures occur
752 mostly on the opposite sides where the shadowing effect is dominant (**Fig. A1b** and d). For the high Sun
753 at SZA 5° , τ failures are almost negligible because the VNIR reflectance observations does not exceed the
754 LUT τ upper limit of 158.48, while there is a small number of occurrences of the r_e too large and r_e too
755 small failures (**Fig. A1a** and c).

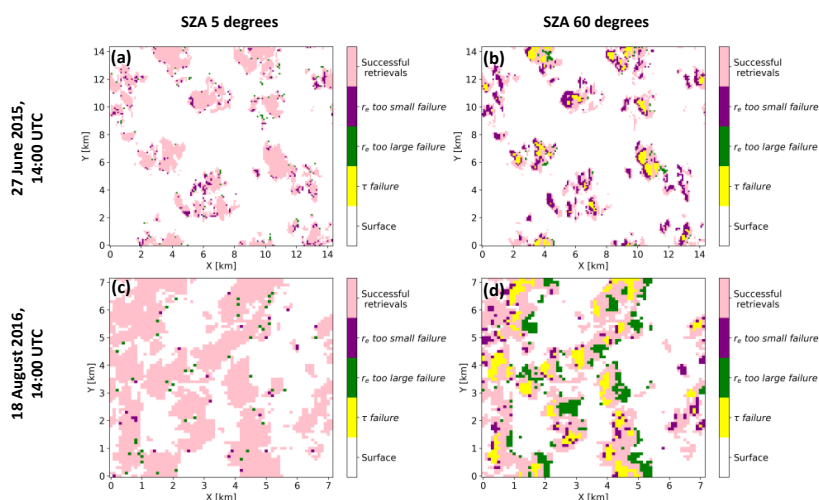
756 It should be noted that when we exclude the failed retrievals from the broadband flux analysis,
757 we keep the total cloud fraction constant. In other words, we scale the broadband flux based on the
758 successful pixels by the ratio of total cloudy to successful pixels such that the effect of cloud fraction
759 reduction is removed from the analysis. The impacts of excluding failed retrievals on the domain-averaged
760 broadband flux can be assessed by comparing the values outside the parentheses with those inside in
761 **Table 4**, and better understood in the light of failed retrieval statistics given in **Table 3**.

762 Results for the 27 June 1D-RT + retrieved clouds experiment at SZA 5° , show that the
763 domain-averaged TOD reflected flux is underestimated by 1.50Wm^{-2} (213.94Wm^{-2} in comparison to
764 215.44Wm^{-2}) when only successful pixels are used to represent the total population of cloudy pixels
765 compared to results which utilize all retrieved cloud pixels in the radiative flux analysis. This is mainly
766 because the dominant type of retrieval failure in this case is the r_e too small failure, accounting for about
767 71% of the failed pixel retrieval statistics (see **Table 3**). Recall that r_e too small failure is mainly a result of
768 illuminating effect and therefore associated pixels appear brighter in 3D RT than 1D RT. As a result,
769 excluding these pixels leads to an underestimate of domain-averaged broadband reflected flux. For the



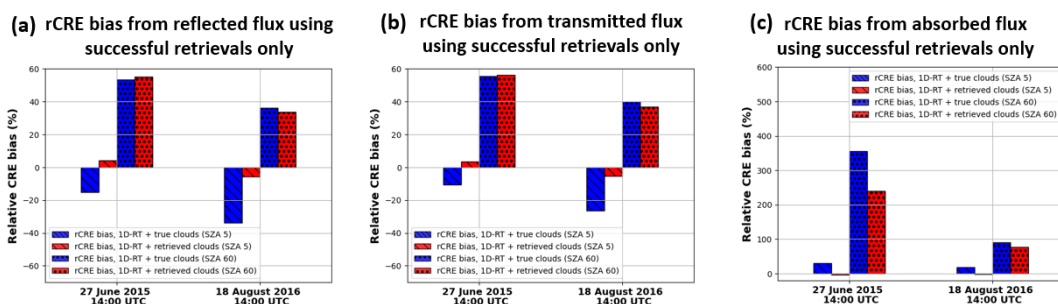
770 same reason, excluding these pixels leads to an overestimation of transmitted flux at the domain bottom
 771 .

772 In contrast to the 27 June case, excluding the failed retrievals in the 18 August case leads to an
 773 overestimation of domain-averaged reflected and underestimation of the surface transmitted flux. This is
 774 probably because the dominant failed retrieval type is the r_e too large which is because of the shadowing
 775 effect. These pixels appear darker from the perspective of TOD and more transmissive from the
 776 perspective of bottom in 3D RT than 1D RT. For comparison purpose, we have also excluded the failed
 777 pixels from the 1D RT + true clouds. Overall, the results are very similar and consistent with those based
 778 on 1D RT + retrieved clouds.



779
 780 Fig A1. Plots of successful and failed retrievals categories for the 27 June 2015 and 18 August 2016 cases at Solar zenith angle 5
 781 degrees (a and c) and Solar zenith angle 60 degrees (b and d).

782 In comparison with the high Sun case, the impacts of failed retrievals on the broadband flux
 783 statistics are much larger in the low Sun SZA 60° case. In both LES cases, the exclusion of failed retrievals
 784 leads to a significant decrease of domain-averaged reflected flux and increase of the transmitted flux. For
 785 example, in the 27 June case, the reflected flux decreased from 134.22 Wm^{-2} when failed pixels are
 786 included to 111.21 Wm^{-2} when they are excluded, which is accompanied by an increase of the
 787 transmitted flux from 419.60 Wm^{-2} to 441.77 Wm^{-2} . A close look at **Table 3** reveals that in both LES
 788 cases, the combination of r_e too small and τ failures accounts for the majority of failed retrievals, 95% in
 789 the case of 27 June and 68% in the 18 August case. As mentioned above, both types of failures are because
 790 of the illuminating effect. Excluding them is expected to cause underestimation of domain-averaged
 791 reflected flux and overestimation of transmitted flux.



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Fig.A2. Relative cloud radiative effect bias computed from the successful only retrievals, top of the domain reflected in (a), surface transmitted in (b) and column absorbed flux in (c) for the two cloud fields.

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805 Author contribution

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811 Competing interest

812

The contact author has declared that none of the authors has any competing interests.

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