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1   **Methodology to determine the coupling of continental clouds with surface from**

2   **lidar and meteorological data**

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15 **Abstract**

16 The states of coupling between clouds and surface or boundary-layer have been

17 investigated much more extensively for marine stratocumulus clouds than for

18 continental low clouds, partly due to more complex thermodynamic structures over

19 land. A manifestation is a lack of robust remote sensing methods to identify coupled

20 and decoupled clouds over land. Here, we have generalized the concept of coupling

21 and decoupling to low clouds over land, based on potential temperature profiles.

22 Furthermore, by using ample measurements from a lidar and a suite of surface

23 meteorological instruments at the U.S. Department of Energy's Atmospheric

24 Radiation Measurement Program's Southern Great Plains site from 1998 to 2019, we

25 have developed a method to simultaneously retrieve the planetary boundary layer

26 (PBL) height (PBLH) and coupled states under cloudy conditions during the daytime.

27 The coupled states derived from lidar show strong consistency with those derived

28 from radiosondes. Retrieving the PBLH under cloudy conditions that has been a

29 persistent problem in lidar remote sensing, is resolved in this study. Our method can

30 lead to high-quality retrievals of the PBLH under cloudy conditions and the

31 determination of cloud coupling states. With the new method, we find that coupled

32 clouds are sensitive to changes in the PBL with a strong diurnal cycle, whereas

33 decoupled clouds and the PBL are weakly related. Since coupled and decoupled

34 clouds have distinct features, our new method offers an advanced tool to separately

35 investigate them in climate systems.

1 Introduction

A large fraction of low clouds is driven by surface fluxes through the conduits of the planetary boundary layer (PBL) over land (e.g., Betts, 2009; Ek and Holtslag, 2004; Golaz et al., 2002; Teixeira and Hogan, 2002; Zheng et al., 2020; Wei et al., 2020; Santanello et al., 2018). This is a coupled cloud-surface system (Cheruy et al., 2014; Zheng & Rosenfeld, 2015; Wu et al., 1998). However, not all low clouds respond to surface forcing. Those clouds without close interactions with the local surface are considered to be in a decoupled state. Given that the PBL is, by definition, the lowest atmospheric layer influenced by the underlying surface (Stull, 1988), to what degree the PBL top overlaps with cloud bases becomes a good criterion to separate coupled and decoupled low clouds.

Conventionally, the "coupled state" of a cloud-topped marine boundary layer implies that the moist conserved variables are vertically well mixed within the PBL (Bretherton and Wyant, 1997; Dong et al., 2015; Zheng & Li, 2019; Zheng et al., 2018). However, such a definition cannot be simply applied to clouds over land since the moist conserved variables typically show considerable variations due to the relatively complex thermodynamics (Driedonks, 1982; Stull, 1988). The definition and the determination methods of the PBL over land also widely differ from those over ocean (Garratt, 1994; Vogelezang & Holtslag, 1996). The concept of coupled and decoupled states is typically used to characterize marine stratocumulus clouds due to their large-scale coverages (Nicholls, 1984). Since stratocumulus only constitutes a relatively small portion of continental clouds (Warren et al., 1986), we attempt to

58   extend the concept of coupling and decoupling to characterize low clouds over land.

59   Following parcel theory, the lifted condensation level (LCL) has been used to

60   diagnose a coupled cloud, based on the distance between the LCL and the cloud base

61   (e.g., Dong et al., 2015; Glenn et al., 2020; Zheng & Rosenfeld, 2015; Zheng et al.,

62   2020). When potential temperature and humidity are uniformly distributed in the

63   vertical, the LCL should be consistent with the cloud base for coupled cases. However,

64   the cloud base for coupled cases can considerably differ from the LCL over land

65   because potential temperature and humidity have large variabilities in the vertical

66   scale within the PBL over land (Guo et al., 2016; Stull, 1988; Su et al., 2017a).

67   Therefore, a robust remote sensing method is still warranted to distinguish coupled

68   and decoupled clouds over land.

69   Since the PBL height (PBLH) is the maximum height directly influenced by

70   surface fluxes, we consider coupling with the PBL equivalent to coupling with the

71   land surface. Thus, we use the PBLH as a critical parameter to diagnose the coupling

72   between clouds and the land surface. The degree of coupling may thus be gauged in

73   terms of quantitative differences between the cloud base and the PBL top. Such

74   differences can be determined in a height coordinate system or in a potential

75   temperature coordinate system (Kasahara, 1974). For this purpose, ground-based lidar

76   has great potential because it can continuously track the development of the PBL

77   (Demoz et al., 2006; Hageli et al., 2000; Sawyer & Li, 2013; Su et al., 2017b) and

78   clouds (Clothiaux et al., 2000; Platt et al., 1994; Zhao et al., 2014) at high temporal

79   and vertical resolutions.

80   By jointly using lidar measurements and meteorological data from the U.S.

81   Department of Energy's Atmospheric Radiation Measurement (ARM) Southern Great

82   Plains (SGP) site (36.6°N, 97.48°W), we attempt to identify coupled and decoupled

83   low clouds during the daytime. Unlike previous studies that use the LCL or

84   radiosonde (RS) data to diagnose coupled clouds (e.g., Dong et al., 2015; Zheng &

85   Rosenfeld, 2015), this study provides the first lidar-based method to automatically

86   determine the coupling and decoupling of low clouds over land at a high temporal

87   resolution.

88   The paper is organized as follows. Section 2 describes the measurements and data.

89   Section 3 describes the new methodology in terms of the definition and

90   implementation. The performance of the method is demonstrated in Section 4, and a

91   summary is presented in Section 5.

92

93   **2   Data Descriptions**

94   *2.1 Radiosonde*

95   RS launches took place at least four times per day at the ARM SGP site, usually

96   at 0030, 0630, 1230, and 1830 local time (LT). Holdridge et al. (2011) provide

97   technical   details   about   the   ARM   RS

98   (https://www.arm.gov/capabilities/instruments/sonde). In this study, we consistently

99   use daylight saving time (Coordinated Universal Time -5 h) as local time throughout

100   the year to avoid inconsistencies between summer and winter. Besides the routine

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101   measurements, there are fewer, but still considerable numbers of RS data obtained at

102   other times of the day (e.g., 0930, 1200, 1300, 1530, and 1900 LT). These

103   supplemental RS samples at other times comprise ~10% of the total number of cases.

104   RS data from 0630–1900 LT are utilized in this study. The vertical resolution of RS

105   data varies according to the rising rate of the balloon, but measurements are generally

106   taken ~10 m apart. We further vertically average the RS data to achieve a vertical

107   resolution of 5 hPa. By using the well-established method developed by Liu and

108   Liang (2010), we retrieved PBLHs over the SGP site from RS measurements. The

109   potential temperature is corrected as the virtual potential temperature, θ_v, using the

110   water vapor mixing ratio [WVMR; $\theta_v = (1 + 0.61\text{WVMR})$]. The virtual potential

111   temperature does not include a correction for the liquid water content profile, as this is

112   challenging to measure in many conditions. Therefore, the virtual potential

113   temperature is not conserved during moist convection. Since we mainly focus on the

114   sub-cloud atmosphere, this is not a serious problem. Moreover, we use scaled RS

115   moisture profiles normalized by the total precipitable water vapor derived from the

116   microwave radiometer (https://www.arm.gov/capabilities/vaps/lssonde, Revercomb et

117   al., 2003).

118

119   *2.2 Micropulse lidar (MPL) system*

120   MPL backscatter profiles were collected at the SGP site from September 1998 to

121   July 2019 with high continuity (Campbell et al., 2002). Technical details and data

122   availability   can   be   found   at   the   website

123 https://www.arm.gov/capabilities/instruments/mpl. The backscatter profiles have a

124 vertical resolution of 30 m. MPL signals have an initial temporal resolution of

125 10–30 s and are averaged every 10 min for this study. Due to the inherent problem of

126 lidar observations, there is a ~0.2-km near-surface "blind zone". Following the

127 standard lidar-data processing, background subtraction, signal saturation and

128 overlapping, after-pulse and range corrections are applied to the raw MPL data

129 (Campbell et al., 2002, 2003). Questionable data are excluded based on the

130 quality-control flags.

131

132 *2.3 Cloud product*

133   The MPL can be used to detect cloud layers based on signal gradients (Platt et al.,

134 1994). Lidar-based methods are accurate for determining the cloud-base height (CBH)

135 but may miss information about the cloud top due to the signal saturation within an

136 optically thick cloud (Clothiaux et al., 2000). Under this condition, the cloud radar

137 provides a better estimation of the cloud-top height (CTH). In this study, we directly

138 use an existing quality-controlled cloud product, CLDTYPE/ARSCL

139 (https://www.arm.gov/capabilities/vaps/cldtype), which combines information from

140 the MPL, ceilometer, and cloud radar to determine the vertical boundaries of clouds

141 (Clothiaux et al., 2000; Flynn et al., 2017). For the lowest cloud base, the best

142 estimation from laser-based techniques (i.e., MPL and ceilometer) is used. The

143 original temporal resolution of the CLDTYPE/ARSCL product is 1 min, averaged to a

144 10-min temporal resolution. To avoid averaging jumps in signal between different

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145   clouds, a cloud is considered to be continuous if its base height varies less than 0.25

146   km between two consecutive profiles.

147

148   **3   Methodology**

149   *3.1 Definition of coupled and decoupled clouds based on thermodynamics*

150   The definition of the state of cloud-surface coupling over land is a critical question.

151   For marine stratocumulus, coupled clouds are identified when the liquid water

152   potential temperature varies less than a certain threshold (i.e., 0.5 K) below the cloud

153   base (Jones et al., 2011). We try to extend the concept of coupling and decoupling to

154   clouds over land. The PBL over land is typically buoyancy driven and controlled by

155   surface fluxes during the daytime. We consider a cloud is in the coupled state when it

156   strongly interacts with the buoyancy fluxes within the PBL.

157   Figure 1 presents the idealized vertical profiles of virtual potential temperature (θ_v)

158   under the clear-sky, coupled cloud, and decoupled cloud. A superadiabatic surface

159   layer exchanges the heat fluxes between the surface and PBL. The outer layer and

160   entrainment zone are turbulently coupled with the surface, and thus, are considered as

161   the coupled regime. Meanwhile, the free atmosphere is considered as the decoupled

162   regime. Theoretically, θ_v is constant in the outer layer, and follows the wet adiabatic

163   lapse rate in the cloud layer. Although the profiles of θ_v in the real atmosphere can

164   largely differ from the idealized profiles, the relative position between the cloud layer

165   and capping inversion of entrainment zone is clear. For the coupled cases, the cloud

166   base is below the capping inversion of entrainment zone. For the decoupled cases, the

167   cloud base is above the capping inversion. Based on this feature, we can use the

168   virtual potential temperature profiles to diagnose the coupling state of low clouds.

169   We first look at several examples of profiles of θ_v and WVMR from the RS

170   (Figure 2). If the CBH is lower than the PBLH, the cloud is affected by turbulence

171   and buoyancy fluxes in the PBL, such as the cases shown in Figure 2a. Note that the

172   PBLH is not an absolute boundary limiting turbulence and buoyancy fluxes. Due to

173   the overshooting of rising air parcels, we use a range to screen the condition of

174   coupled clouds. As shown in Figure 2b, even when the CBH is slightly above the

175   PBLH, WVMR and θ_v are still relatively consistent between the cloud layer and the

176   PBL and show large step signals at the cloud top.

177   Figure 2c-d shows a clear inversion layer between the cloud base and the PBL top,

178   and the difference in θ_v between the CBH and the PBLH ($\Delta\theta_v$) is relatively large.

179   Such a notable inversion layer prevents the buoyancy fluxes within the PBL from

180   reaching the cloud base, leading to the decoupling between the cloud and the PBL.

181   Overall, whether there is a clear inversion between the cloud base and the PBL top is

182   the key factor in determining coupling and decoupling. In this aspect, $\Delta\theta_v$ is the key

183   factor. In Figure 2, $\Delta\theta_v$ for coupled cases (a-c) is -0.32 K and 0.31 K, respectively,

184   and $\Delta\theta_v$ for decoupled cases (d-e) is 1.47 K and 5.0 K, respectively.

185   Therefore, instead of giving a height range to limit the differences between CBH

186   and PBLH, we consider using the differences in θ_v between CBH and PBLH ($\Delta\theta_v =$

187    $\theta_v^{CBH} - \theta_v^{PBLH}$) to determine the threshold for distinguishing coupled and decoupled

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188   clouds. For decoupled cases, the cloud base is above the capping inversion of

189   entrainment zone. There is a notable inversion in θ_v between PBL top and decoupled

190   cloud base. Thus, we identify the cases satisfying $\Delta\theta_v > \delta_s$ as being in a decoupled

191   state. Correspondingly, we identify the cases satisfying $\Delta\theta_v < \delta_s$ as being in a

192   coupled state. We set the range of CBH to between 0 and 4 km and excluded cases of

193   deep convection (i.e., CBH < 4 km and CTH > 6.5 km).

194   For identical cases, θ_v is constant within the PBL. Under this assumption, δ_s

195   can be set as 0. However, there are certain variations in θ_v within the PBL, which

196   can cause inversions with relatively small magnitudes between the cloud base and

197   PBL top. Figure 3a presents the inversion strength in θ_v within PBL during the

198   daytime. The inversions near surface or across the PBL top are excluded. Besides the

199   capping inversion and surface inversion, the inversion strength within PBL is

200   typically below 1K. Therefore, we set δ_s as 1 K, which is the same as the criterion

201   for determining stable or convective conditions (Liu & Liang, 2010). Furthermore, we

202   demonstrate the probability density function (PDF) of $\Delta\theta_v$ for the low cloud cases.

203   Coupled and decoupled clouds are classified by the threshold of δ_s (1 K). Through

204   the development of PBL, boundary layer clouds frequently occur in the entrainment

205   zone, and form a coupled cloud-PBL system. For such coupled systems, θ_v at cloud

206   top and PBL top is highly consistent for the majority of cases. Thus, the PDF of $\Delta\theta_v$

207   shows significantly high values for the range of -2 K to 0.5 K in the coupled regime.

208   Meanwhile, the PDF of $\Delta\theta_v$ is evenly distributed in the decoupled regime. Since we

209   only analyze low clouds, the PDF of $\Delta\theta_v$ slowly decrease when $\Delta\theta_v$ is above 10 K.

210   Based on the variations in θ_v within PBL, we set δ_s as 1 K. However, it should

211   note that it is not an absolute value. A similar threshold of 0.5 K has been used for

212   marine stratocumulus (Jones et al., 2011; Dong et al., 2015). Comparing to the marine

213   condition, θ_v show greater variabilities over land. Hence, the threshold is

214   correspondingly larger. On the other hand, since the threshold of 1 K is in the low

215   PDF regime (Figure 3b), the small changes in this value would not notably affect the

216   identifications. Specifically, a 0.1 K difference in δ_s will lead to a 0.5% difference in

217   the identification of coupled cloud.

218   Following the traditional definition of buoyancy forces (Wallace & Hobbs, 2006),

219   we further integrate the buoyancy forces within the lowest 1 km ($- g \int_0^{1km} \frac{1}{\theta} \frac{\Delta\theta}{\Delta z} dz$).

220   Figure 4 shows the relationships between CBH and buoyancy forces in the lower

221   atmosphere for $\Delta\theta_v < \delta_s$ (coupled state) and $\Delta\theta_v > \delta_s$ (decoupled state). In terms of

222   the responses to buoyancy forces, dramatic differences between coupled and

223   decoupled clouds are seen. Following our previous study (Su et al., 2018), we use the

224   inverse fitting ($f(x) = \frac{A}{x} + B$) with consideration of density to establish the

225   relationship between CBH and buoyancy forces. The magnitude of the correlation

226   coefficient ($R^\dagger$) is designed to measure the degree to which the data fit an inverse

227   relationship. For a coupled cloud, changes in CBH to variable buoyancy forces mostly

228   follow an inverse function. For the coupled cases, strong buoyancy forces are

229   associated with a thick PBL and high CBH. Since a decoupled cloud occurs in the free

230   atmosphere, the CBH of a decoupled cloud has a very weak linkage with the

231   buoyancy forces.

232

3.2 Lidar-based method to identify coupled and decoupled clouds

3.2.1 Method description

Given the rapid change in clouds over land, RS observations have limitations when it comes to tracking cloud development due to the coarse temporal resolution and drifting of the balloon. We thus further developed a lidar-based method to identify the coupled states of clouds based on our new algorithm for retrieving the PBLH that can better track the diurnal variations in PBLH than conventional lidar-based approaches (Su et al., 2020). We adapted this algorithm for retrieving the PBLH and developed a new scheme to deal with cloudy conditions. The principles behind the PBLH algorithm are stated next for completeness.

Our new PBLH algorithm can retrieve the PBL variability from the MPL under Different Thermo-Dynamic Stability (thus named the DTDS algorithm) conditions, taking into account the vertical coherence and temporal continuity of the PBLH. First, we identify the local maximum positions (LMPs; range: 0.25–4 km) in profiles of the wavelet covariance transform function derived from lidar backscatter (Brooks, 2003). These LMPs are the potential positions of the PBLH. We can use the PBLH derived from morning RS soundings as the starting point. Without morning RS soundings, the algorithm can still work well, with the lowest LMPs selected as the starting point, which reduces by 0.02–0.05 the correlation coefficient between MPL-derived and RS-derived PBLHs (Su et al., 2020).

253   To ensure good continuity, we select the closest LMP to the earlier position of the

254   PBLH. Different stages of PBL development are considered. DTDS-derived PBLHs

255   likely increase during the growth stage and decrease during the decaying stage, but

256   the algorithm is also able to identify decreases during the growth stage or increases

257   during the decaying stage based on the selection scheme described by Su et al. (2020).

258   There are multiple step signals in the backscatter profiles when complex aerosol

259   structures (e.g., the residual layer) are present, leading to multiple LMPs. Based on

260   temporal continuity, we select the appropriate LMP as the position of the PBL top.

261   However, PBLH retrievals still suffer from relatively low accuracies under stable

262   conditions because of the weak vertical mixing and residual layer.

263   Clouds induce strong step signals in the lidar backscatter, further considerably

264   affecting PBLH retrievals. Su et al. (2020) only considered cases where the low cloud

265   top coincided with the previous PBL top, excluding other low-cloud cases (> 60% of

266   all low-cloud cases). Here, we specifically consider coupled and decoupled states of

267   low clouds. Due to the MPL's ~0.2-km blind zone, we only analyze the PBLH and

268   CBH above 0.2 km. Figure 5 presents the flow chart describing the updated DTDS

269   algorithm. In particular, we jointly use PBL development and the LCL to diagnose the

270   states of coupling or decoupling. In ideal situations, LCL, PBLH, and CBH are highly

271   consistent with each other for coupled clouds. But for real conditions, we only require

272   that either the LCL or the PBLH coincides with the CBH for identifying coupled cases,

273   with another parameter serving as an additional constraint. Specifically, a coupled

274   cloud needs to occur within a certain range of LCL and the previous position of the

275   PBL top. For the DTDS algorithm, five empirical parameters are used, including A_1,

276   A_2, A_3, A_4, A_5. As listed in the Table 1, $A_1 - A_5$ are set as 0.7, 0.2, 0.15, 1.35, and

277   1.1, respectively. A cloud at time i is identified as being in the coupled state if the

278   CBH is less than $[H(i-1) + 0.2$ km $(A_2)]$ and $[\text{LCL} + 0.7$ km $(A_1)]$. A cloud is also

279   considered to be in a coupled state if the CBH is coincident with the LCL within 0.15

280   km (A_3), and the CBH is less than $[H(i-1) + 0.7$ km $(A_1)]$, where $H(i-1)$

281   represents the PBLH at time $(i-1)$. The LCL is calculated from surface

282   meteorological data (relative humidity, temperature, pressure) at the SGP site based

283   on an exact expression (Romps, 2017).

284   After determining the coupling or decoupling state of a cloud, we retrieve $H(i)$

285   (i.e., PBLH at time i) based on the cloud state. For decoupled cases, we use the same

286   strategy for a clear sky to retrieve the PBLH. Based on the selection scheme in the

287   DTDS algorithm, the LMP below the CBH is selected as $H(i)$. For coupled cases, we

288   jointly use CBH and CTH to determine PBLH. During the warm season, shallow

289   cumulus often occurs in the upper part of the PBL with strong surface heating, so the

290   CTH can be regarded as the PBLH (Stull, 1988; Wallace & Hobbs, 2006). Under this

291   condition, the CTH coincides with the previous PBL top. Therefore, if $[\text{CTH} <$

292    $PBLH_{30min} + 0.2\ km\ (A_2)]$, we set $H(i)$ equal to the minimum between CTH and the

293   product A_4*CBH, where $PBLH_{30min}$ is the average value of the PBLH within 30

294   min of the prior time i. Since this does not work for the active cloud (i.e. a cloud that

295   reaches its level of free convection), we set $H(i) = A_5$CBH, if $[\text{CTH} \geqslant PBLH_{30min} +$

296    $0.2\ km\ (A_2)]$. After retrieving $H(i)$, we consider that the cloud above the PBLH is

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297   still coupled if [CBH $< H(i) + 0.2 \, km \, (A_2)$]. Moreover, we added an upper limit for

298   all PBLH retrievals. If [$H(i) >$ LCL $+ 0.7 \, km \, (A_1)$], we adjust $H(i)$ as the maximum

299   LMP below the LCL. The new DTDS method combines lidar measurements and

300   surface meteorological observations and can simultaneously retrieve the PBLH and

301   cloud states.

302

303   *3.2.2 Selection of empirical parameters*

304   The states of coupling and decoupling are diagnostic parameters rather than

305   explicit expressions. Similar to the other methods for retrieving PBLH (e.g., Brooks,

306   2003; Liu & Liang, 2010), multiple empirical parameters are used to determine PBLH.

307   Here we discuss the selection of empirical parameters in the algorithm.

308   Note that we used the CTH and A_4*CBH as the upper limits for PBLH retrievals

309   in the DTDS algorithm. For coupled cases, these two limits are generally close to or

310   above the position of the PBL top. Only 2% (3%) of total cases meet the condition

311   that the RS-derived PBLH is 0.25 km higher than the CTH (A_4*CBH). Section 4

312   presents the detailed relationships between CBH, CTH, and PBLH. In the DTDS

313   method, CTH serves as the upper limit for PBLH under the condition of coupled

314   shallow cumulus.

315   Similar to previous studies, we can also use the LCL as the standard to identify

316   coupled clouds (Dong et al., 2015; Zheng & Rosenfeld, 2015). We assume a cloud is

317   coupled if |CBH − LCL| < some criteria. Figure 6a shows the commission errors and

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318   omission errors for different criteria. Here, the commission error is calculated as the

319   percentage of decoupled clouds misidentified as coupled clouds. The commission

320   error can also be called a "false positive", as the former is a common term for

321   describing the nature of an error in identification. The omission error is calculated as

322   the percentage of coupled clouds that have not been identified under this criterion. By

323   using the LCL, we can obtain a relatively low commission error if the criterion is less

324   than 0.15 km and a relatively low omission error if the criterion is greater than 0.7 km.

325   Thus, we set A_1 and A_3 as 0.7 and 0.15 in the DTDS method to exclude and to

326   select cases of coupled clouds. We can also use the RS-derived PBLH as the criterion

327   (Figure 6b). Despite the coarse temporal resolution, the RS-derived PBLH can be a

328   good criterion to use to distinguish between coupling and decoupling. If we consider a

329   coupled cloud as a cloud where (CBH < RS-derived PBLH + 0.2 km), both

330   commission and omission errors are ~5%. Therefore, we primarily use [PBLH+0.2

331   km (A_2)] in the DTDS method to identify coupled and decoupled regimes.

332   Moreover, we test the sensitivity of selecting these empirical parameters. Figure 7

333   presents the commission errors and omission errors in the identifications of coupled

334   clouds for selecting different values of empirical parameters. Among these parameters,

335   A_2 is the critical one, which would notably affect the identification results. In general,

336   A_2 determine the maximum differences between PBLH and CBH for coupled cases.

337   If [CBH-PBLH > A_2], we consider the cloud is under the decoupled state. Thus, the

338   identification method is quite sensitive to A_2. Selecting a low value of A_2 would

339   neglect many coupled cases, which leads to a high omission error. Meanwhile,

340 selecting a high value of A_2 would misclassify many coupled cases, which leads to a

341 high commission error. After a trail and error, A_2 is set as 0.2 km to balance the

342 omission and commission errors. The selections for other parameters are not sensitive

343 for the coupled cloud identifications. We can choose them from a reasonable range.

344 As a by-product of this method, we also pay attentions to the PBLH retrievals

345 under cloudy conditions. Figure 8 presents the mean absolute biases and correlation

346 coefficients between PBLH derived from lidar and radiosonde for selecting different

347 values of empirical parameters. To match the scope of this study, we only analyze the

348 low cloud conditions. For retrieving PBLH under cloudy conditions, A_2 is the critical

349 parameter. The variations in correlation coefficients under different values of

350 empirical parameters are small with a range of 0.81-0.82. However, the absolute

351 biases can considerably differ under different values of A_5. In general, A_5 represents

352 the ratio between CBH and PBLH under coupled conditions. If A_5 is above 1.1,

353 PBLH retrievals under cloudy conditions are overestimated. We set A_5 as 1.1 to

354 achieve a relatively low biases and a relatively high correlation coefficient at the same

355 time. For other parameters, the selections from reasonable ranges would not notably

356 affect the PBLH retrievals.

357 In short, selections of these empirical parameters are based on the overall

358 relationship between cloud and PBL under the coupled and decoupled states. In our

359 method, the selection of A_2 is critical for the identifications of coupled clouds, while

360 the selection of A_5 is critical for the PBLH retrievals under cloudy conditions. The

361 selections of other parameters are not sensitive.

362

4 Results

364 Figure 9 illustrates four examples of PBLH retrievals and cloud states derived

365 from the DTDS algorithm for 27 October 2011, 31 July 2002, 19 March 2000, and 1

366 May 2012. Figure 9a depicts coupled shallow cumulus occurring at noontime at the

367 PBL top. With a weak surface flux of ~200 W m^{-2}, this shallow cumulus cloud

368 appeared for less than an hour. Figure 9b shows a developed coupled cumulus cloud.

369 With a strong surface flux of ~500 W m^{-2}, this coupled cloud continuously developed

370 during the daytime. Figure 9c presents the case of a daylong coupled cloud. After the

371 passage of a frontal system that day, stratocumulus occurred during the morning with

372 a cloud thickness of 0.5 km. Through the development of the PBL, the thick

373 stratocumulus cloud was broken up by the strong turbulences, transforming into

374 shallow cumulus clouds. Figure 9d shows the case of an active coupled cloud, which

375 is generally associated with a large amount of convective available potential energy.

376 Even though coupled clouds can differ in appearance and variability throughout the

377 day, the common feature is the coherent variation between the cloud base and the PBL

378 top. The LCL is a relevant parameter and can differ from the PBLH and the CBH for

379 some coupled cases (e.g., Figure 9b-c).

380 The identification accuracy, or disparity between different methods, are evaluated

381 in terms of the selected criteria, for which the identification method based on $\Delta\theta_v$ is

382 regarded as the "truth", as described in Section 3.1. Hereafter, all results are analyzed

383 for the period of 1000–1900 LT, so early-morning data are not used. The commission

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384 error is 10.1%, and the omission error is 6.8% for the DTDS method. Note that

385 lidar-based PBLH methods generally suffer from relatively low accuracy under stable

386 atmospheric conditions. Following Liu and Liang (2010), we identified stable PBLs

387 from RS measurements. Since coupled clouds are driven by relatively strong

388 buoyancy fluxes, only 1% of total cases of coupled clouds occurred under stable PBL

389 conditions during the study period (0700–1900 LT). Therefore, the relatively low

390 accuracy for stable PBLs is not a major problem in this study.

391   Figure 6 also compares the accuracy between the DTDS and LCL methods.

392 Based on the LCL alone, we cannot choose an appropriate criterion to achieve a lower

393 commission error and omission error simultaneously. Thus, we do not use the LCL as

394 the single standard to detect the coupling and decoupling of low clouds in our study.

395 As diagnostic parameters, different methods inevitably produce different results

396 regarding coupling and decoupling. Although we consider the method based on $\Delta\theta_v$

397 as the standard, it still suffers from uncertainties arising from balloon drifting. From

398 this perspective, it is hard to conclude which method is the best. Since it determines

399 the PBLH based on aerosol backscattering, the lidar-based method may be more

400 representative of the coupling between a cloud and the aerosol layer near the surface

401 when clear skies occur, at least during a short window of time.

402   Figure 10a-b presents the occurrence frequencies of the CBH and the CTH at

403 different heights. Despite the same variation ranges, clouds are mostly coupled if the

404 CBH is lower than 1 km, while decoupled clouds dominate if the CBH is higher than

405 3 km. Figure 10c-d shows the changes in the coupled fraction (ratio of coupled cases

406   to total cases) with different CBHs and CTHs. The coupled fraction is about 90% if

407   the CBH is lower than 1 km and decreases to 2% for CBHs above 3 km. Although the

408   CBHs for coupled cases are generally less than 3 km, CTHs for coupled cases can be

409   much higher. Coupled clouds still account for around 10% of the cases with CTHs

410   above 6 km.

411   Figure 11 shows scatter plots between CBH, CTH, PBLH, and LCL for coupled

412   and decoupled clouds. For coupled clouds, there is a generally strong correlation

413   between CBH, LCL, and PBLH, contrary to the weak relationships of decoupled cases.

414   The relationship between CTH and RS-derived PBLH is complicated. For shallow

415   cumulus clouds, their tops can be considered as PBL tops for the coupled state, while

416   the cloud top is considerably above the position of the PBL top for active cumulus

417   clouds. We also note that the accuracy of CTH retrievals is generally lower than the

418   accuracy of CBH retrievals (Clothiaux et al., 2000). As CTH is not a criterion for

419   cloud coupling, the accuracy of CTH would not affect the identification of coupled

420   cloud, but may affect the PBLH retrievals for the coupled cloud cases.

421   For coupled clouds, the DTDS-derived PBLH relies on the cloud position,

422   showing a strong correlation with the RS-derived PBLH with a correlation coefficient

423   (R) of ~0.9 (Figure 11d), since the position of the coupled cloud serves as a good

424   reference for determining the PBLH. For decoupled cases, the correlation between the

425   DTDS-derived PBLH and the RS-derived PBLH is generally good (R = 0.73) but

426   worse than the correlation for coupled cases. As pointed out in previous studies (Chu

427   et al., 2019; Hageli et al., 2000; Lewis et al., 2013; Su et al., 2017b), it has been a

428   persistent problem to retrieve the PBLH under cloudy conditions since the large

429   backscatter and step signals from cloud interference would be excluded to avoid

430   interfering with the retrievals. After identifying the coupling or decoupling of clouds,

431   the PBLH can be successfully retrieved under cloudy conditions.

432   Compared to the clear-sky cases discussed in previous studies (e.g., Chu et al.,

433   2019; Yang et al., 2017), the DTDS-derived PBLH shows a much higher correlation

434   with RS-derived PBLH for coupled cloud cases and has a similar R as the RS-derived

435   PBLH for decoupled cloud cases. Moreover, due to the different definitions of the

436   PBLH and aerosol stratification within the PBL, there are always considerable

437   differences between lidar- and RS-derived PBLHs, which cannot be eliminated by a

438   specific algorithm (Chu et al., 2019; Su et al., 2020).

439

440   **5   Summary**

441   In this study, we proposed a novel method for distinguishing between coupled

442   and decoupled low clouds over land. Based on the understanding of PBL processes

443   and quantitative analyses, we developed a lidar-based method (DTDS) to identify the

444   coupling state of low clouds over the SGP site. In practice, we identified a coupled

445   cloud when the position of the cloud base was generally close to or lower than the

446   previous position of the PBL top, with the LCL serving as an additional restriction.

447   Compared to using the LCL alone, the coupled states identified by the DTDS method

448   show better consistency with the results derived from radiosondes, with about 10%

449   differences between the lidar-based retrievals and radiosonde results.

450   Not only coupled state, the method also can provide high-quality retrievals of the

451   PBLH under cloudy conditions. A long-lasting problem with lidar-retrieval of PBLH

452   is either incapability of retrieval or large uncertainties induced by the occurrence of

453   low clouds (e.g., Chu et al., 2019; Hageli et al., 2000; Lewis et al., 2013), we address

454   this issue by separately considering the coupled and decoupled of low clouds.

455   Specifically, in coupled conditions, the position of the coupled cloud serves as a good

456   reference for identifying the PBLH. In decoupled conditions, the large backscatter and

457   step signals from clouds would be excluded to avoid interfering with the retrievals.

458   With our method, cloudy conditions are well handled.

459   With the new method, we study the difference of cloud-PBL interactions in

460   coupled and decoupled conditions. In contrast to the sensitive responses of coupled

461   clouds to changes in the PBLH and buoyancy, the decoupled clouds and the PBLH are

462   weakly related. Due to their different relationships with the PBL, a robust

463   distinguishment between the coupled and decoupled low clouds is critical for further

464   investigating the coupled land-atmosphere system and aerosol-cloud interactions. Our

465   methodology paves a solid ground for such pursuits.

466

467   *Code/Data availability.* All these datasets are publicly available at the ARM archive

468   https://adc.arm.gov/discovery/#/results/site_code::sgp. The products developed in this

469   study, i.e., cloud states and the PBLH, are currently available upon request from the

470   lead author (tianning@umd.edu) and are expected to be added to the ARM archive in

471   the near future. The code of this method is available upon request from the lead author

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472   (tianning@umd.edu).

473

474   *Author contribution.* T.S. and Z.L. conceptualized this study. T.S. carried out the

475   analysis, with comments from other co-authors. T.S., Y.Z., and Z.L. interpreted the

476   data and wrote the manuscript.

477

478   *Competing interests.* The authors declare that they have no conflict of interest.

479

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663 **Tables**

664 **Table 1.** List of parameters in the flow chart of DTDS (Figure 5). These parameters

665 are related with three factors, including LCL, PBLH, CBH. The sensitivity of

666 selection of these parameters is presented. The detailed impacts of variations in these

667 parameters on the retrievals of cloud coupling and PBLH are presented in Figure 7

668 and Figure 8, respectively.

669

670

	Unit	Related factors	Value	Sensitivity (coupled states)	Sensitivity (PBLH)
A_1	km	LCL / PBLH	0.7	Low	Low
A_2	km	PBLH	0.2	**High**	Low
A_3	km	LCL	0.15	Low	Low
A_4	dimensionless	CBH	1.35	Low	Low
A_5	dimensionless	CBH	1.1	Low	**High**

671   **Figures**
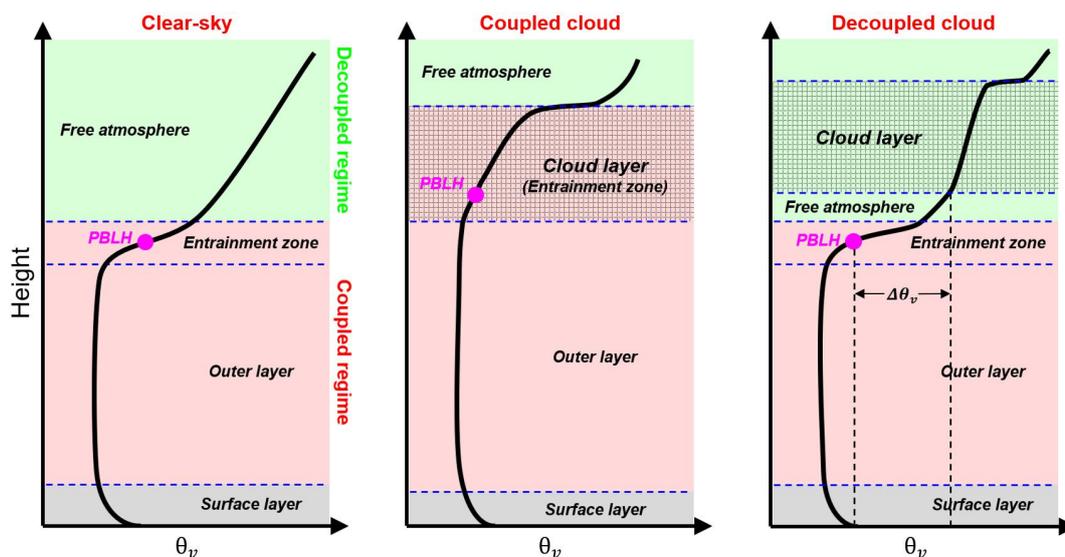
Figure 1. Idealized vertical profiles of virtual potential temperature (θ_v) under the clear-sky, coupled cloud, and decoupled cloud over land. The surface layer, outer layer entrainment zone, and free atmosphere are divided by the blue dash lines. The cloudy layer is marked as the shaded area, and PBLH is marked as the pink point. Red and green zones indicate the coupled and decoupled regime, respectively. Elements (e.g., turbulence, heat fluxes, cloud) in the coupled regime are directly affected by the PBL processes, while these elements are not directly affected by the PBL processes in the decoupled regime. For the coupled cases, the cloud base is below the capping inversion of entrainment zone. For the decoupled cases, the cloud base is above the capping inversion.
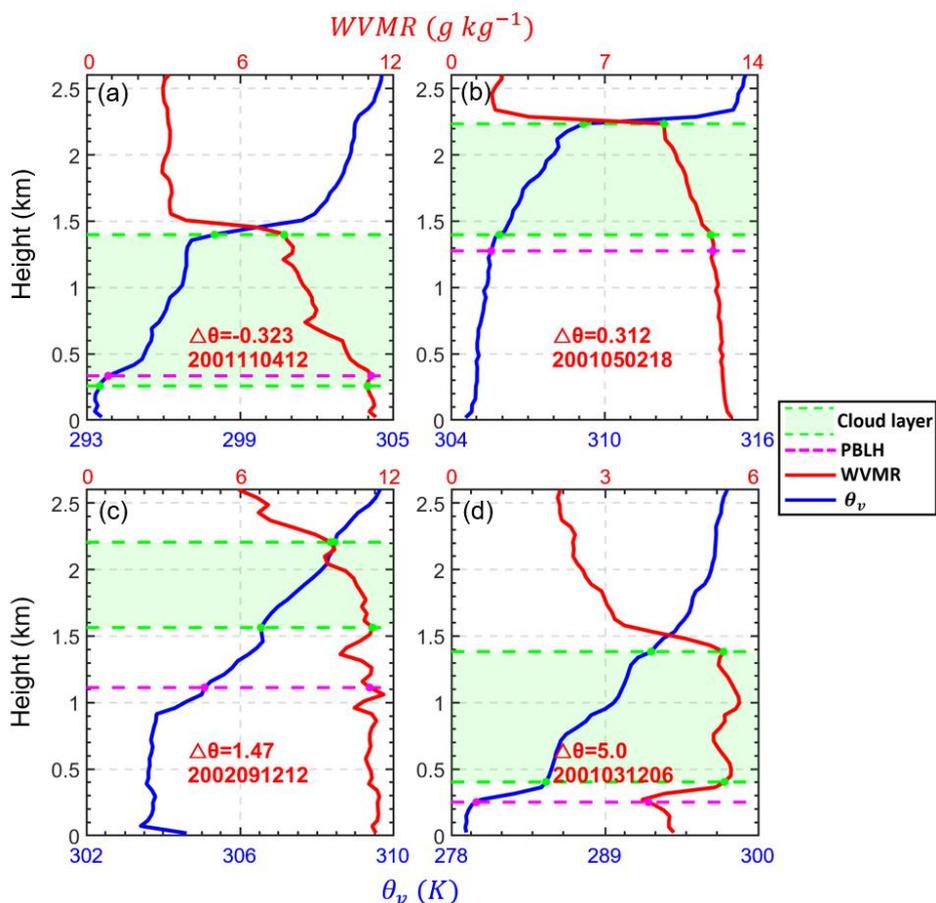
Figure 2. Virtual potential temperature (θ_v, red lines) and water vapor mixing ratio

(WVMR, blue lines) profiles from radiosonde (RS) over the Southern Great Plains

site for different cases. The differences in virtual potential temperature between the

cloud base and the planetary boundary layer (PBL) top are expressed as $\Delta\theta_v$ ($\theta_v^{CBH} -$

θ_v^{PBLH}) . The time of each radiosonde launch is marked in each panel as

"YYYYMMDDHH", where YYYY, MM, DD, and HH indicates the year, month, day,

and local time, respectively. Green regions are cloud layers, and green dashed lines

indicate their boundaries. PBL heights are marked as dashed pink lines.
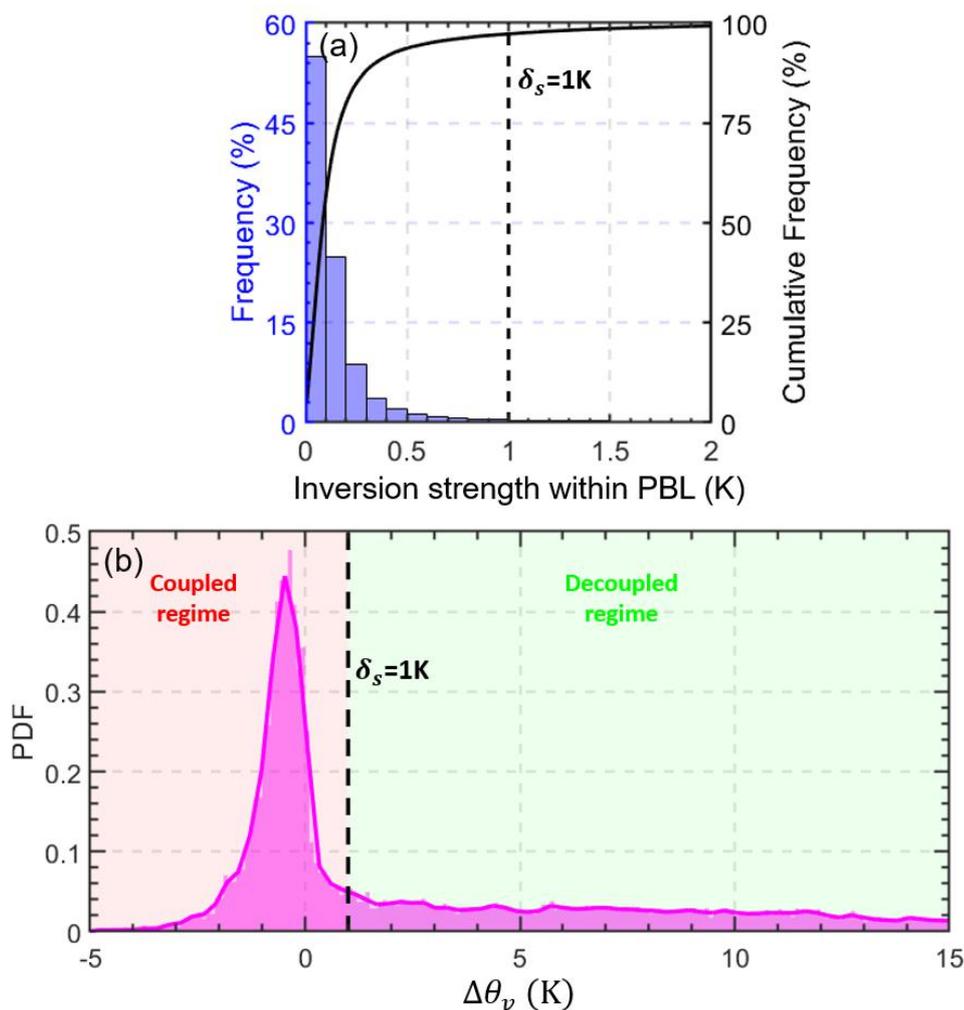
Figure 3. (a) Blue bars represent the inversion strength of θ_v within the PBL. The inversion strength is derived from the radiosonde during daytime (0800-1900LT). The inversions near surface or across PBL top are excluded. The black solid line represents cumulative frequency. (b) Pink area represents the probability density function (PDF) of the differences in the virtual potential temperature between cloud-base height (CBH) and PBLH ($\Delta\theta_v = \theta_v^{CBH} - \theta_v^{PBLH}$). By using a threshold of δ_s (1 K), coupled and decoupled regimes are classified.
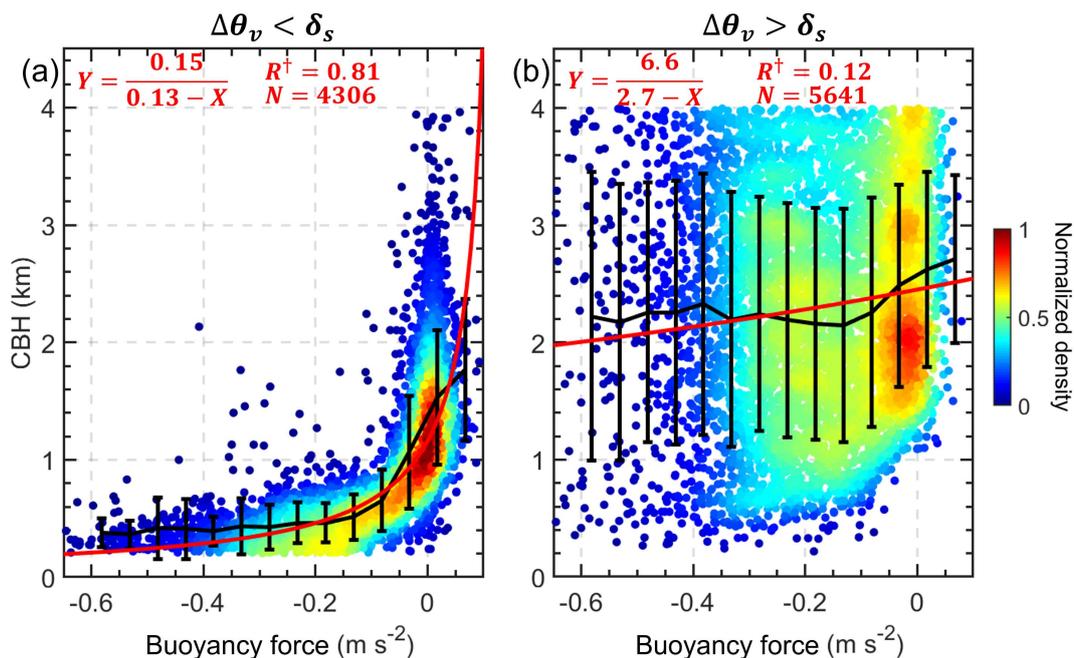
Figure 4. The relationships between cloud-base height (CBH) and buoyancy forces in the lower atmosphere (0–1 km) for (a) $\Delta\theta_v < \delta_s$ and (b) $\Delta\theta_v > \delta_s$. $\Delta\theta_v$ represents the differences in the virtual potential temperature (θ_v) between CBH and PBLH. Black lines represent the mean values in each bin, and the whiskers indicate the standard deviations. Red lines represent the inverse fits. The fitting functions and number of samples (N) are given in each panel, along with the correlation coefficient ($R^\dagger$) for the inverse fit.

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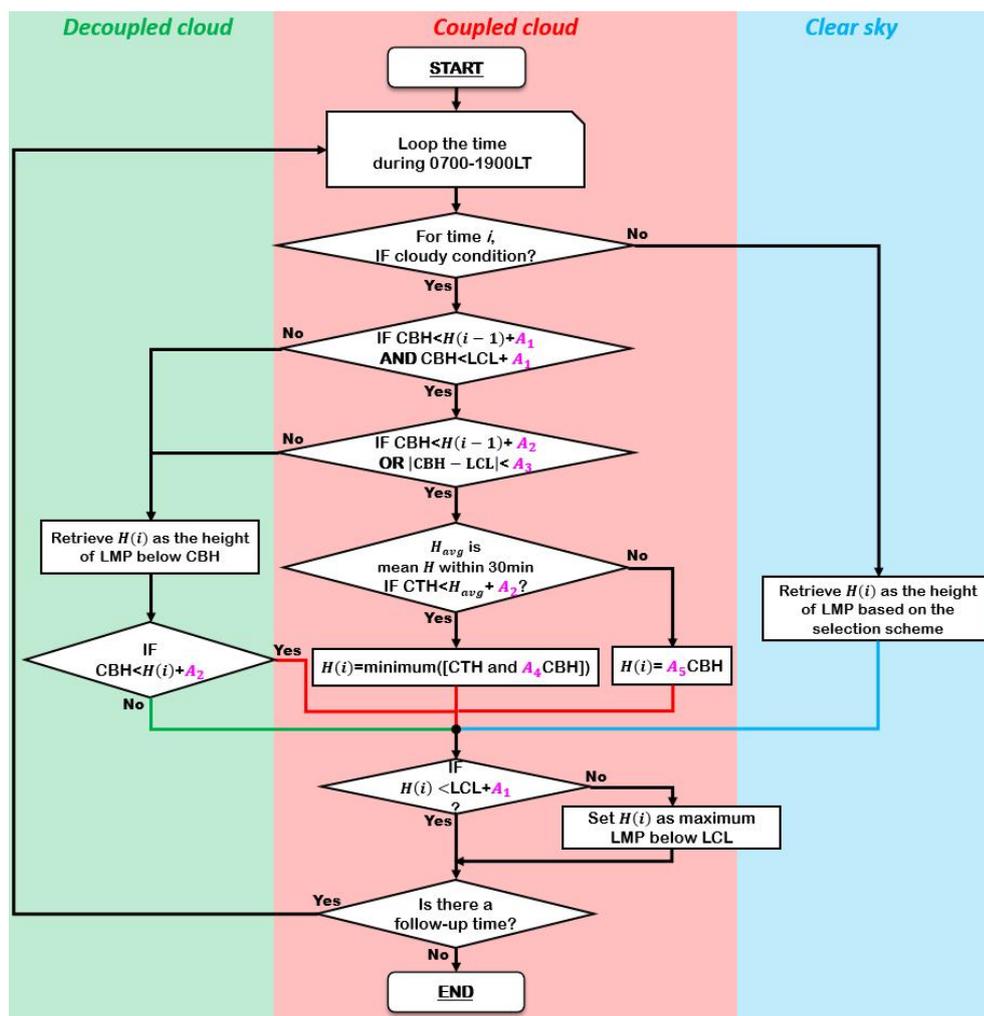
Figure 5. The flow chart of the updated DTDS algorithm. In this diagram, $H(i)$ is the retrieved planetary boundary layer height (PBLH) at time i. CBH and CTH represent the base and top heights, respectively, of the lowest cloud at time i. The PBLH part for selecting the suitable local maximum position (LMP) follows Su et al. (2020), and a detailed scheme for identifying a coupled cloud is added to the DTDS algorithm. LCL stands for lifted condensation level. Five empirical parameters (A_1, A_2, A_3, A_4, A_5) are set as 0.7, 0.2, 0.15, 1.35, 1.1, respectively.

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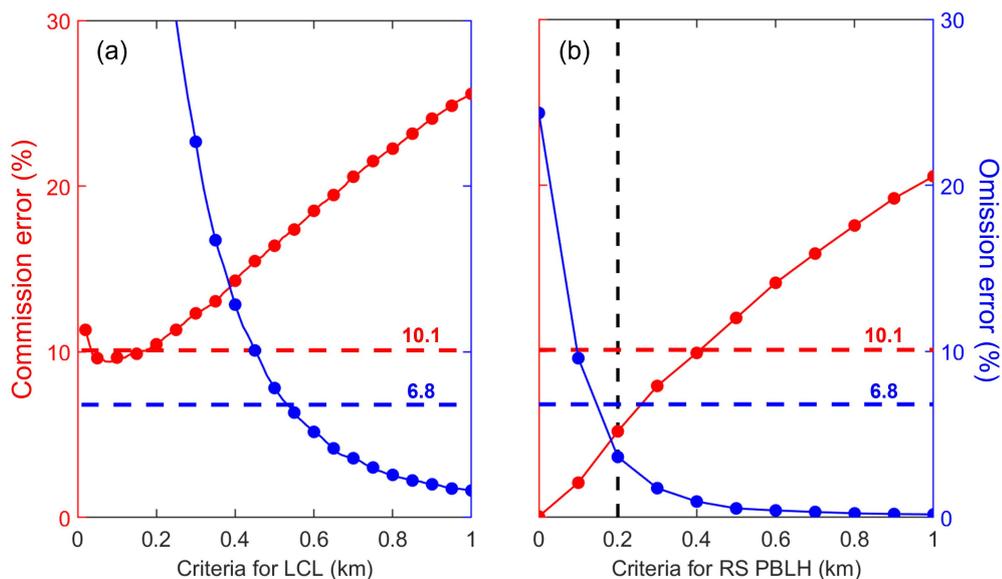
Figure 6. Commission errors and omission errors of coupled cloud identifications (a)

for different criteria for the lifted condensation level (LCL) and (b) for different

criteria for the planetary boundary layer height (PBLH). "Criteria for LCL" means

coupled clouds are identified if $|CBH - LCL| <$ Criteria for LCL. Similarly, "Criteria

for RS PBLH" means coupled clouds are identified if $CBH - RS\ PBLH <$ Criteria

for RS PBLH. The red and blue dashed lines indicate the commission and omission

errors, respectively, for the DTDS algorithm. CBH stands for cloud-base height, and

RS stands for radiosonde.

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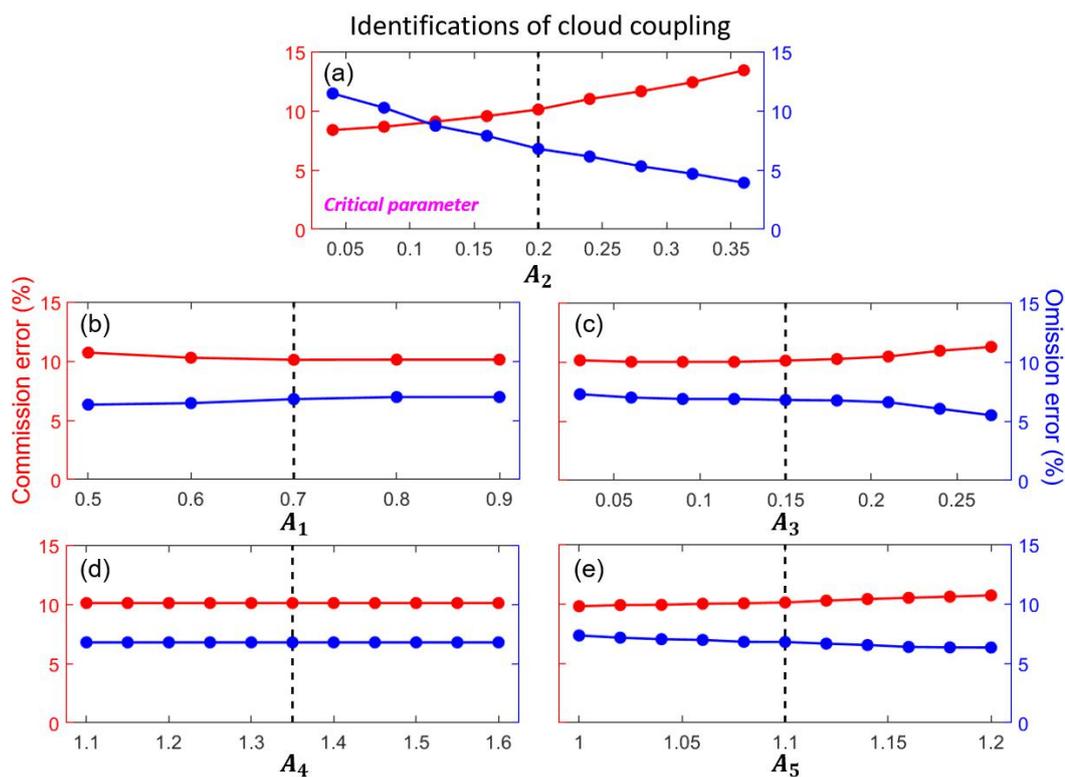
Figure 7. Commission errors (red line) and omission errors (blue line) of coupled

cloud identifications for selecting different values of empirical parameters

(A_1, A_2, A_3, A_4, A_5) in the DTDS algorithm. Black dash lines indicate the default

values. For each test, one parameter is variable, while other parameters are set as

default values. For identifications of cloud coupling, A_2 is the critical parameter.

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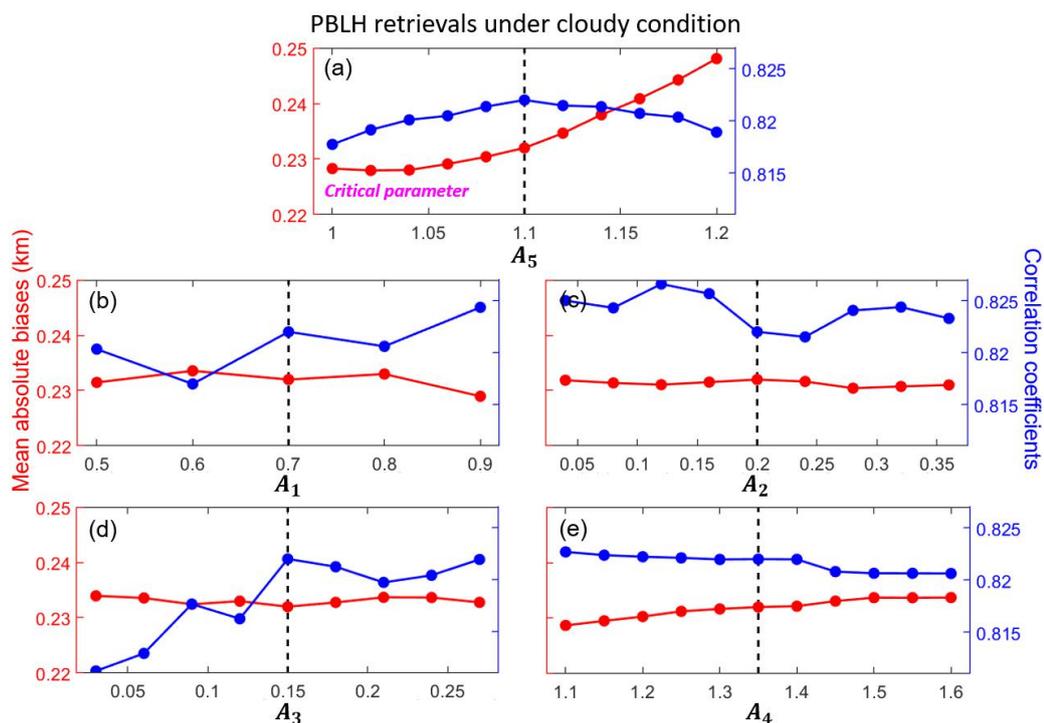
Figure 8. Red lines indicate the mean absolute biases between PBLH derived from lidar and radiosonde for selecting different values of empirical parameters (A_1, A_2, A_3, A_4, A_5) in the DTDS algorithm. Here, we only analyze the low cloud cases. Blue lines indicate the corresponding correlation coefficients between PBLH derived from lidar and radiosonde. Black dash lines indicate the default values. For each test, one parameter is variable, while other parameters are set as default values. For PBLH retrievals under cloudy conditions, A_5 is the critical parameter.
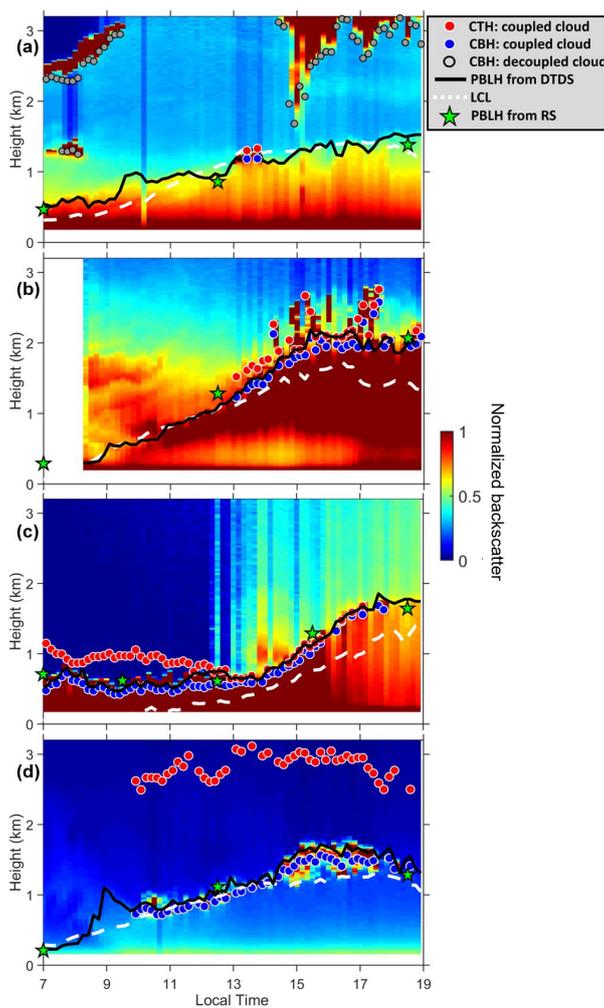
Figure 9. Daily backscatter profiles: (a) short-lived coupled cloud, (b) developed coupled cloud, (c) daylong coupled cloud, and (d) active coupled cloud. Backscatter is normalized to a range of 0–1 in arbitrary units. Red dots and blue dots indicate cloud-top heights (CTHs) and cloud-base heights (CBHs) of coupled clouds. Grey dots mark CBHs for decoupled clouds. Black lines and green stars mark the planetary boundary layer height (PBLH) retrieved from the DTDS algorithm and from radiosonde (RS) soundings, respectively. White dashed lines represent lifted
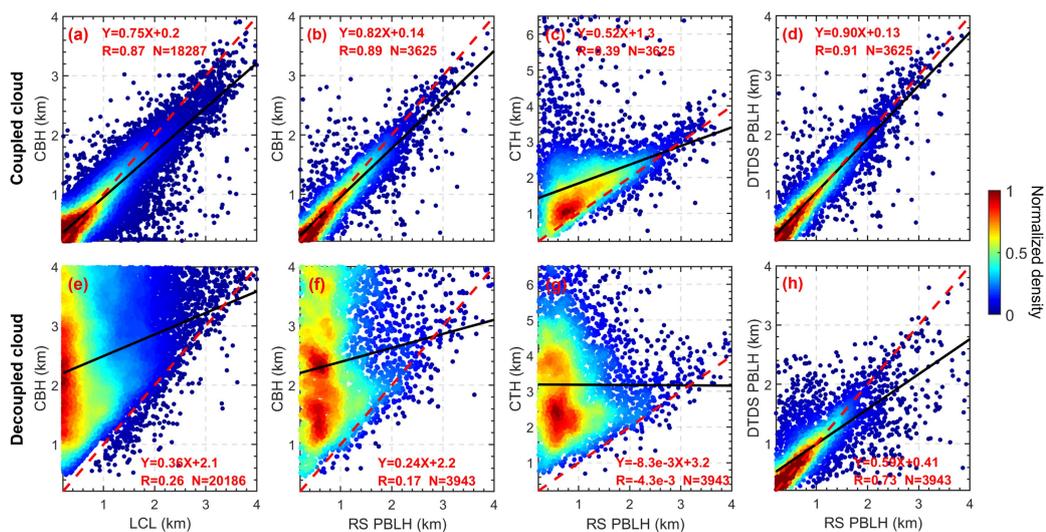
753   condensation levels (LCLs).
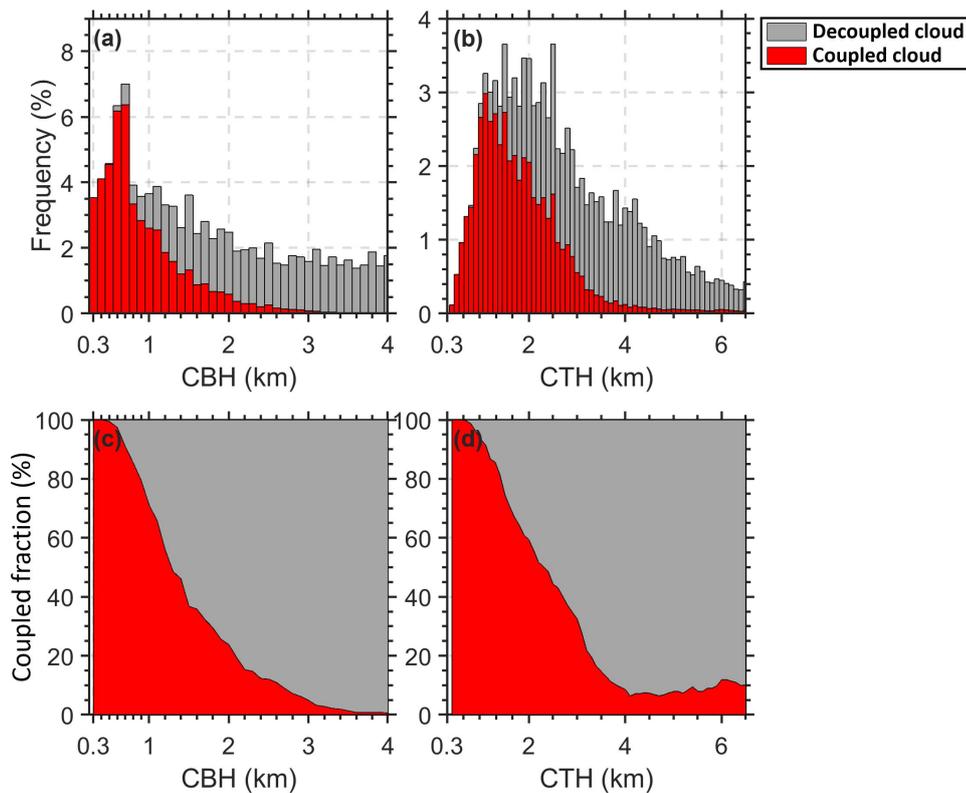
Figure 10. The height-dependent occurrence frequencies of (a) the cloud-base height

(CBH) and (b) the cloud-top height (CTH) for coupled clouds (red bars) and

decoupled clouds (grey bars). The relative occurrence frequencies of (c) the CBH and

(d) the CTH for coupled clouds (red area) and decoupled clouds (grey area).

Figure 11. The relationships between (a) LCL and CBH, (b) CBH and RS-derived PBLH, (c) CTH and RS-derived PBLH for coupled clouds, and (d) DTDS-derived PBLH and RS-derived PBLH. Panels (e-h) are similar to panels (a-d) but for decoupled clouds. Black lines represent the linear regressions. The linear fitting functions, correlation coefficients (R), and sampling numbers (N) are given in each panel.