

1 **Seasonally Transported Aerosol Layers over Southeast Atlantic**
2 **are Closer to Underlying Clouds than Previously Reported**

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27 **Abstract:**

28 From June to October, low-level clouds in the Southeast (SE) Atlantic often underlie
29 seasonal aerosol layers transported from African continent. Previously, the Cloud-
30 Aerosol Lidar and Infrared Pathfinder Satellite Observation (CALIPSO) 532 nm lidar
31 observations have been used to estimate the relative vertical location of the above-cloud
32 aerosols (ACA) to the underlying clouds. Here, we show new observations from NASA's
33 Cloud-Aerosol Transport System (CATS) lidar. Two seasons of CATS 1064 nm
34 observations reveal that the bottom of the ACA layer is much lower than previously
35 estimated based on CALIPSO 532nm observations. For about 60% of CATS nighttime
36 ACA scenes, the aerosol layer base is within 360 m distance to the top of the underlying
37 cloud. Our results are important for future studies of the microphysical indirect and semi-
38 direct effects of ACA in the SE Atlantic region.
39

40 **1. Introduction**

41 Every year from about June to October over the southeast (SE) Atlantic, the prevailing
42 easterly winds in the free troposphere often transport the smoke and pollution aerosols
43 from the African continent to the west, over the ocean where extensive marine boundary
44 layer (MBL) clouds persist for most of the year [Adebiyi and Zuidema, 2016]. This leads
45 to a near-persistent seasonal biomass burning aerosol layer over MBL clouds in SE
46 Atlantic [Devasthale and Thomas, 2011; Zhang *et al.*, 2016].

47 As summarized in Yu and Zhang [2013] instruments onboard NASA's A-train satellite
48 constellation provide valuable observations of the aerosol layer and underlying clouds. In
49 particular, the lidar on the space-borne mission CALIPSO provides unique observations
50 of the vertical distribution of the aerosol layer that have been widely used to characterize
51 the aerosol layer above cloud over SE Atlantic [Chand *et al.*, 2008; Yu *et al.*, 2010;
52 Devasthale and Thomas, 2011; Meyer *et al.*, 2013] and assess its impacts on the radiation
53 budget [Chand *et al.*, 2009; Zhang *et al.*, 2016].
54

55 The seasonally transported SE Atlantic aerosol layer can influence the regional radiative
56 energy budget through the direct radiative effect (DRE) [Chand *et al.*, 2009; Zhang *et al.*,
57 2016]. The absorption by aerosol layer can also influence the thermodynamical structure
58 of lower atmosphere and in turn change cloud field, which is known as the semi-direct
59 effect [Johnson *et al.*, 2004; Wilcox, 2010; Sakaeda *et al.*, 2011; Wilcox, 2012]. The sign
60 and magnitude of the semi-direct effect are strongly dependent on the vertical distribution
61 of aerosol with respect to the underlying clouds [Johnson *et al.*, 2004]. In addition to
62 DRE and semi-direct effect, the aerosol particles could be entrained into the clouds and
63 activated as cloud condensation nuclei, giving rise to the so-called aerosol indirect effects
64 [Costantino and Bréon, 2010; 2013; Painemal *et al.*, 2015]. Intuitively, the closer the
65 bottom of the aerosol layer gets to the top of underlying cloud, the more likely the aerosol
66 particles are entrained into the cloud. Previous studies have used the 532 nm observations
67 from the CALIPSO lidar to estimate the distance from the aerosol layer bottom to the
68 cloud top (referred to hereafter as AB2CT distance for short). Costantino & Bréon [2010]
69 show that 84% of the time the AB2CT distance in SE Atlantic is larger than 250m.
70 Devasthale and Thomas [2011] found that in 0° to 30°S region, 90-95% of above-cloud-

71 aerosol cases has an AB2CT distance greater than 100m. Yu et al. [2010] derived the
72 average AB2CT of 1700 m over a two-year period in SE Atlantic. These analyses based
73 on CALIOP 532 nm observations seem to indicate that the seasonal aerosol layer in SE
74 Atlantic is well separated from the underlying clouds and thus the aerosol indirect effects
75 may be secondary in comparison to the aerosol direct and semi-direct effects (e.g.,
76 [Sakaeda et al., 2011]).

77
78 It is known that the CALIOP 532 nm based layer detection often misses the lowest
79 boundary of a thick aerosol layer, thereby biasing the bottom of the aerosol layer too
80 high. This may be especially problematic for daytime observations [Meyer et al., 2013].
81 Recently, several novel remote sensing techniques have been developed to retrieve the
82 AOD (Aerosol Optical Depth) of above-cloud absorbing aerosol layers from passive
83 sensors (e.g. [Waquet et al., 2009; Torres et al., 2011; Meyer et al., 2015]). In addition,
84 an alternative lidar method has been developed for CALIOP, utilizing signals from the
85 underlying cloud instead of the attenuated backscatter profile [Hu et al., 2007; Liu et al.,
86 2015]. When compared with the retrievals from passive sensors and the alternative
87 CALIOP algorithm, the operational 532nm CALIOP AOD retrievals are systematically
88 biased low by 26% on average [Liu et al., 2015], and can be up to a factor of 5 lower
89 [Jethva et al., 2014]. A likely explanation for this bias is that the strong aerosol
90 attenuation at 532 nm by the upper portion of the aerosol layer together with the small
91 backscatter cross section of the aerosol particles, substantially weakens the attenuated
92 backscatter signal from the lower part of the aerosol layer to a level under the detection
93 threshold of CALIOP [Kacenelenbogen et al., 2011; Torres et al., 2013; Jethva et al.,
94 2014; Liu et al., 2015]. This laser attenuation issue leads to an overestimation of the
95 aerosol layer bottom height (too high), an underestimation of the physical thickness of the
96 aerosol layer (too thin), and thereby an underestimation of AOD (too small).

97
98 In this study, we seek to shed new light on the vertical distribution of the SE Atlantic
99 absorbing aerosol layer with respect to the underlying clouds using observations from
100 NASA's CATS mission. Because of instrument and algorithm differences, CATS ACA
101 retrieval suffers much less from the laser saturation-induced bias than CALIOP 532nm
102 algorithm. We do a comparative analysis of CATS and CALIOP retrievals in the SE
103 Atlantic region for two recent biomass burning seasons (2015 and 2016). As shown in the
104 letter, the CATS 1064nm observations suggest that bottom of the ACA layer is much
105 lower, and therefore closer to underlying cloud top, than previously estimated based on
106 CALIOP 532nm observations. Our results are important for future studies of the
107 microphysical indirect, as well as the semi-direct, effects of ACA on underlying clouds.

108 **2. Data**

109 The occurrence frequency of above-cloud-aerosol in the SE Atlantic (20W to 20E; 30S to
110 10N) is highest during July-to-October (JASO) with the peak during August-September
111 [Zhang et al., 2016]. In this study, we focus on the two biomass burning seasons (JASO)
112 of 2015 and 2016 so that we can directly compare CALIPSO and CATS (Figure 1).

113 **2.1. CALIOP**

114 The lidar instrument onboard the CALIPSO mission, which has an orbital height of ~700
115 km, is the Cloud-Aerosol Lidar with Orthogonal Polarization (CALIOP). CALIOP

116 directly measures the range-resolved total (particulate plus molecular) attenuated
117 backscatter signal at two wavelengths, 532nm and 1064nm, using analog detection. In
118 addition to the total attenuated backscatter, CALIOP also measures two orthogonal
119 polarized components of the 532nm-backscatter signal [Winker *et al.*, 2009]. The
120 accuracy of the CALIOP Level-2 (L2) data products (aerosol type, particulate backscatter
121 and extinction coefficient, optical depth) is dependent on the accurate detection of cloud
122 and aerosol layers.

123

124 Uniform cloud and aerosol layer detection and cloud-aerosol discrimination (CAD)
125 techniques are challenging due to the complexity of atmospheric scenes encountered. The
126 current version CALIOP selective, iterated boundary location (SIBYL) algorithm uses
127 the 532nm total attenuated backscattered signals to determine boundaries of cloud and
128 aerosol layers, with a typical vertical resolution of 30 m [Vaughan *et al.*, 2009]. The
129 SIBYL scheme detects atmospheric features by iteratively comparing horizontally
130 averaged CALIOP 532 nm total attenuated backscatter profiles at multiple horizontal
131 resolutions. The CALIOP CAD algorithm is a multidimensional probability distribution
132 function (PDF) technique [Liu *et al.*, 2004; 2009] based on statistical differences of
133 several cloud and aerosol properties (e.g., layer-integrated 532nm attenuated backscatter,
134 layer-integrated backscatter color ratio, etc.). Previous studies have shown the SIBYL
135 and CAD algorithms perform well for cirrus clouds and several aerosol types [McGill *et*
136 *al.*, 2007; Yorks *et al.*, 2011; Burton *et al.*, 2013].

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138 2.2. CATS

139 CATS is an elastic backscatter lidar employing photon counting detection and two high-
140 repetition rate lasers that operate at 532 and 1064nm [McGill *et al.*, 2015] that has been
141 operating on the ISS since February 2015. The ISS orbit, which is at an altitude of ~415
142 km and a 51-degree inclination, allows CATS to observe locations at different local times
143 each overpass (~60 days to complete full diurnal cycle) with roughly a three-day repeat
144 cycle.

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146 The CATS layer detection algorithm is a threshold-based layer detection method that is
147 nearly identical to the CALIOP-SIBYL technique with four distinct differences, namely
148 the use of 60 m vertical resolution, a single horizontal spatial resolution (5km), the use of
149 the 1064nm wavelength rather than 532nm, and a technique to identify clouds embedded
150 within aerosol layers [Yorks *et al.*, 2015]. The CATS L2 Operational (L2O) CAD
151 algorithm is a multidimensional PDF technique like the CALIOP one [Yorks *et al.* 2015],
152 but uses the layer-integrated attenuated backscatter at 1064 nm and other variables such
153 as layer mid-temperature and layer thickness instead of the layer-integrated backscatter
154 color ratio due to the unreliable 532 nm data in Mode 7.2. The use of a single horizontal
155 spatial resolution in the CATS algorithm misses optically thin cirrus clouds and aerosols
156 during the daytime in the CATS L2O Version 1-05 data products, though it performs well
157 during nighttime observations. Future versions of CATS L2O data products will include
158 layer detection at 60 km, but since Version 1-05 is used in this study, CATS daytime data
159 was excluded.

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161 For above-cloud aerosol (ACA), the more relevant difference between the algorithms is
162 the preferred wavelength for atmospheric layer detection. The current CALIOP-SIBYL
163 primarily uses 532 nm because it has higher signal-to-noise ratios (SNR) and lower
164 minimum detectable backscatter (MDB,weakest aerosol backscatter coefficient that can
165 be detected) than the CALIOP 1064 nm data resulting in more accurate uniform cloud
166 and aerosol layer detection [Vaughan *et al.*, 2009]. The CATS layer detection algorithm
167 uses the 1064 nm attenuated scattering ratio because the CATS 532 nm data in Mode 7.2
168 is extremely noisy and the 1064 nm MDB is orders of magnitude lower [Yorks *et al.*,
169 2016]. For ACA detection specifically, the 1064 nm wavelength is preferred over the 532
170 nm wavelength for layer detection. The aerosol signal at 1064 nm has sixteen times less
171 molecular contamination compared to 532 nm. As discussed in Section 1, the 532 nm
172 backscatter signal may be insensitive to the entire vertical extent of absorbing aerosol
173 layers. Because aerosol extinction is usually smaller at 1064 nm than 532 nm, and the
174 CATS 1064nm backscatter signal is very robust, the vertical extent of absorbing aerosol
175 layers is fully captured from CATS 1064 nm backscatter profiles. It is worth mentioning
176 that the current CATS operational algorithm uses $AB2CT < 360$ m as the threshold to
177 detect the clouds embedded within aerosol layers (CEAL) [Yorks *et al.* 2017]. When
178 $AB2CT < 360$, the ACA and the cloud below is merged and identified a CEAL case.

179

180 The detectability of the aerosol layer base using 532 and 1064 nm is demonstrated in
181 Figure 1. CATS and CALIPSO passed over the same ACA layer over the SE Atlantic on
182 06 August 2016, although the differing orbits of the ISS and CALIPSO mean that the two
183 curtains do not align exactly. There is a 0.1-1.0 km gap between cloud top and aerosol
184 base in the attenuated total backscatter and vertical feature mask based on CALIOP 532
185 nm data. In contrast, CATS 1064 nm observation finds the aerosol plume to extend all the
186 way to the cloud top, which is also confirmed by the CALIOP 1064nm attenuated
187 backscatter observation. The example clearly demonstrates the advantage of 1064nm
188 over 532 nm-based layer detection technique for identifying the bottom of thick smoke
189 layers. Although CALIOP also has the 1064 nm observation, it has not yet been utilized
190 in the current operational algorithm. Note that the differences between CALIOP and
191 CATS observations shown below are mainly due to the use of different wavelength (i.e.,
192 532nm vs. 1064nm) for layer detection. At the moment of writing, the CALIPSO
193 operational product team is planning to make more use of the 1064nm observations in
194 their operational layer detection algorithm, which could significantly improve its
195 retrievals for thick aerosol layers like the example in Figure 1.

196

197 **3. Results**

198 We have used the following criteria to identify ACA columns in both CALIOP and
199 CATS layer products: (1) the cloud layer product identifies liquid phase cloud at the top
200 layer of the profile; (2) the aerosol layer product identifies at least one layer of aerosol in
201 the profile; (3) the base height of at least one aerosol layer is higher than the top of the
202 highest cloud layer. In the SE Atlantic region, most ACA cases are simple, with only one
203 aerosol layer on top of single-layer MBL clouds. After the identification of ACA
204 columns, we compute the $AB2CT$ by calculating the difference between the minimum
205 aerosol base height which is greater than maximum cloud top height and the maximum
206 cloud top height. For CALIOP, we derived the ACA and cloud statistics for both daytime

207 and nighttime conditions (though daytime and nighttime statistics are computed
208 separately). The CATS results are only for nighttime since its aerosol retrieval does not
209 perform well during daytime at the fixed 5 km horizontal resolution as discussed above.

210

211 Figure 2 (first row) shows the multi-year (2015-2016) SE Atlantic JASO Cloud Fraction
212 (CF), defined as $CF = N_{cloudy}/N_{total}$ in $2^{\circ} \times 2^{\circ}$ grid boxes where N_{cloudy} is the number
213 of cloudy columns and N_{total} is the number of total columns. Because we are interested
214 in aerosol above low-level MBL clouds, ACA frequency (ACA_F) is shown in the
215 second row of Figure 2 is defined as $ACA_F = N_{ACA}/N_{cloudy}$ where N_{ACA} is the number
216 of ACA columns. Among the three datasets, CATS nighttime observations identify the
217 highest ACA occurrence frequency, with domain averaged ACA_F around 0.24.
218 CALIOP daytime observations have the lowest ACA occurrence frequency, with domain
219 averaged ACA_F only around 0.17. The CALIOP nighttime observations are comparable
220 to the CATS nighttime observations (domain average ACA_F \sim 0.23). Some differences
221 between the three datasets may have physical explanations. For example, CALIOP
222 observes a larger CF during nighttime than during daytime, which is likely a result of the
223 strong cloud diurnal cycle in the SE Atlantic region [Min and Zhang, 2014]. The other
224 differences may stem from algorithm and instrument differences. For example, the lower
225 ACA_F using daytime CALIOP might be an artifact due to the impact of background
226 solar noise on the lidar retrieval [Liu et al., 2015].

227

228 Overall, the results in Figure 2 suggest that, despite some minor differences, CALIOP
229 and CATS observe similar geographical patterns of ACA in the SE Atlantic. We now
230 focus on the vertical distribution of aerosol and cloud from the two instruments. Figure 3
231 shows the two-year (2015-2016) mean aerosol layer base height (top row), cloud layer
232 top height (middle row) and AB2CT distance (bottom row) of ACA over the SE Atlantic
233 region during JASO from CALIOP and CATS. While the magnitudes differ, cloud top
234 heights from all three datasets show a similar pattern, lowest off the coast of Namibia
235 (near 20S and 10E) and gradually increasing along the northwest direction to about 2km
236 around 5S and 15W. In contrast to the similarity of cloud top height, the mean ACA base
237 height from the three datasets show significant differences. ACA base height from
238 daytime CALIOP observations is much higher than nighttime CALIOP, which is in turn
239 higher than nighttime CATS. As a result, the AB2CT distance from nighttime CATS is
240 below 500m in most of the SE Atlantic region, suggesting that the aerosol layer extends
241 close to the cloud top. On the other hand, a clear separation between aerosol base and
242 cloud top during both daytime and nighttime is implied by the CALIOP data, a likely
243 result of the abovementioned CALIOP ACA layer detection issues.

244

245 We analyzed the AB2CT distances from the three observations further in Figure 4. Here,
246 we show the Cumulative Density Function (CDF) of the AB2CT distance for the
247 sampling-masked ACA cases of Fig. 3. According to CATS nighttime 1064 nm
248 observations (red curve), about 60% of ACA cases are identified as CEAL (i.e.,
249 AB2CT<360m), in contrast to only 15% and 6% occurrence of such cases in CALIOP
250 532nm nighttime (blue curve) and daytime (green curve) observations, respectively.
251 Moreover, 82% and 64% of ACA cases have AB2CT>1 km according to the daytime and

252 nighttime CALIOP 532nm observations, respectively, in contrast to 22% according to
253 CATS observations.

254

255 Figure 5 shows meridionally-averaged daytime (a) and nighttime CALIOP (b) 532nm,
256 and nighttime CATS 1064nm (c) observations of ACA top (dashed red line) and bottom
257 (solid red line) height, cloud top height (blue line), and the fraction of ACA cases with
258 $AB2CT < 360m$ (black line). Also shown are one standard deviation variability for ACA
259 top (red error bars), ACA base (light red shades) and cloud top (light blue shades). All
260 three observations show nearly the same top of aerosol layer, just below 4km. The cloud
261 top heights are also similar in all three observations, rising from 1km near the coast
262 westward to about 1.5-2.0 km at 19W. Daytime CALIOP observes slightly higher cloud
263 top height (domain average 1.39km) compared to nighttime (domain average 1.33km).
264 Among all the observations, the CATS detects the highest cloud top height (domain
265 average 1.60km) among all three data sets. In contrast to aerosol top and cloud top
266 heights, ACA base heights are substantially different among the three data sets. The
267 CALIOP nighttime product (Figure 5b) gives domain-averaged ACA base height at
268 2.63km; daytime CALPSO retrievals (Figure 5a) are even higher. Nighttime CATS 1064
269 nm (Figure 5c), however, observes a significantly lower ACA domain-averaged base
270 height around 2km.

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272 Even after considering one standard deviation variability, there is still a clear separation
273 between the ACA base and cloud top in both the daytime (Figure 5a) and nighttime
274 (Figure 5b) CALIOP retrievals, confirmed by the small values of $AB2CT < 360m$
275 throughout the domain. With CATS (Figure 5c), however, there is clear evidence that the
276 ACA base and cloud top are in much closer proximity than is implied by CALIOP 532nm
277 observation, as the $AB2CT < 360m$ is mostly around 60%.

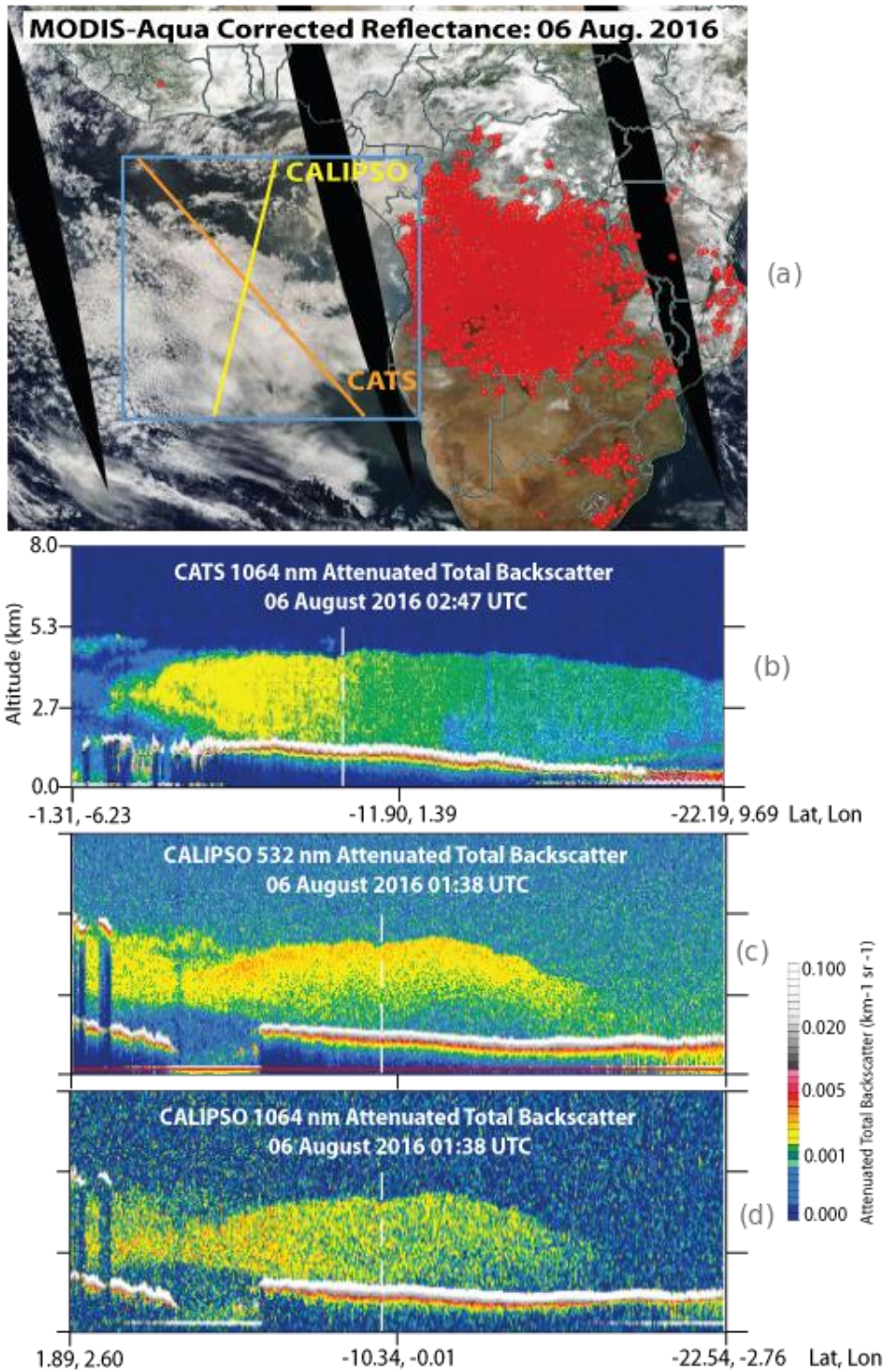
278 4. Summary and Discussion

279 The microphysical indirect effects of the seasonal transported aerosols in the SE Atlantic
280 are often overlooked in the literature. This is partly because CALIOP's 532nm-based
281 operational layer detection algorithm often detects the aerosol layer bottom too high and
282 thereby suggests that the above-cloud aerosol layer is well separated from the underlying
283 clouds. The newly launched CATS mission provides a new dataset of the vertical
284 distribution of aerosol and clouds. Several instrument and algorithm advantages of
285 CATS, chiefly among which is the primary use of 1064 nm for layer detection, allows it
286 to better identify the full vertical extend of the SE Atlantic ACA layer than CALIOP
287 532nm product. We have compared the current CATS and CALPSO products during
288 JASO of 2015 and 2016 over the SE Atlantic. The CF, ACA_F and cloud top
289 geographical patterns from the two instruments agree well. However, CATS 1064nm
290 observes the ACA layer bottom height much lower and much closer to the underlying
291 cloud top than CALIOP 532nm does. According to CATS, about 60% of the ACA cases
292 have an $AB2CT < 360m$, in contrast to the 15% and 6% based on CALIOP nighttime and
293 daytime 532nm observations, respectively.

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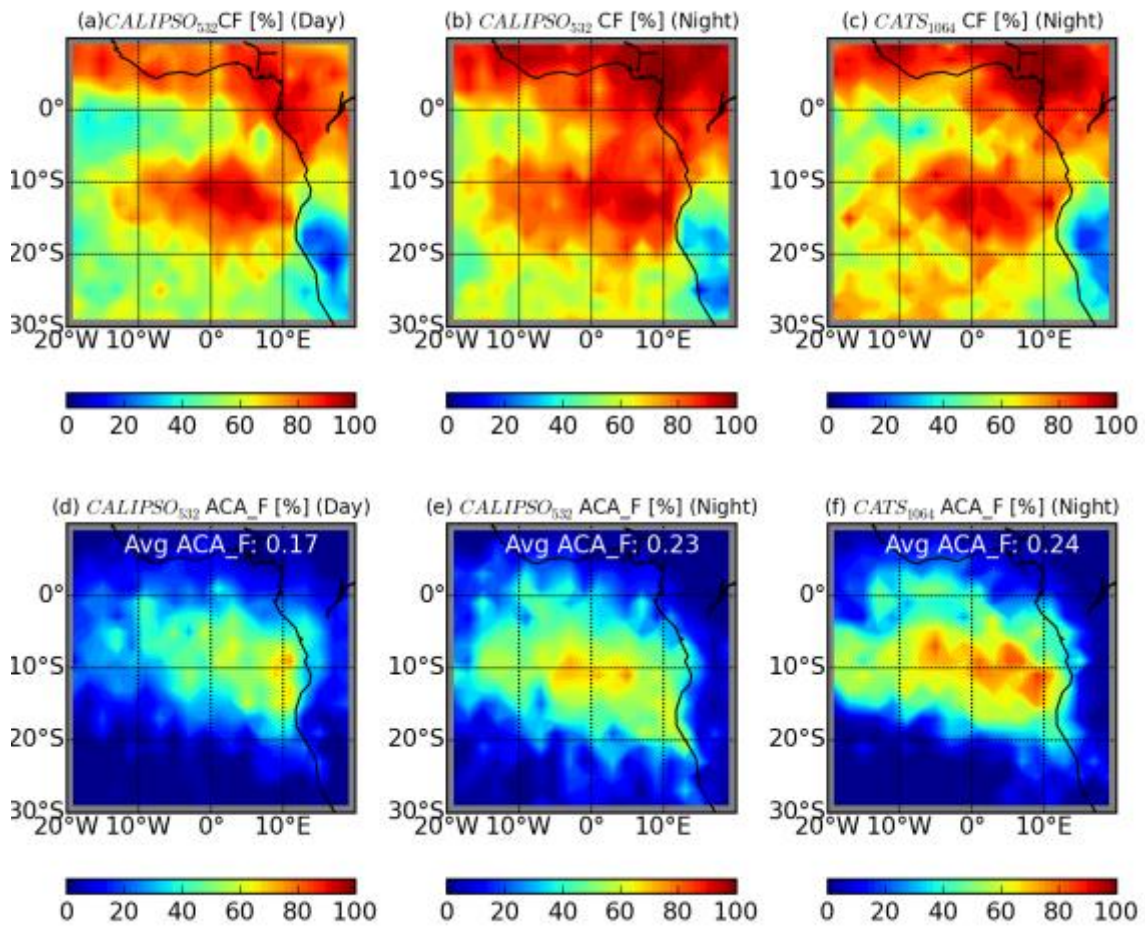
295 Our study provides direct evidence that space-based lidar layer detection at 1064 nm is
296 more representative of the true ACA scene compared to 532 nm. More importantly, our
297 study suggests that the occurrence of aerosol entrainment into clouds might be much

298 more frequent than previously thought based on CALIOP 532nm observations. This
299 implies that the microphysical indirect effects could be an important mechanism through
300 which the transported aerosol influences the clouds and radiation in SE Atlantic region.
301 Finally, an accurate measurement of the vertical distribution of aerosols would also help
302 us better understand the semi-direct effects of the smoke aerosols.
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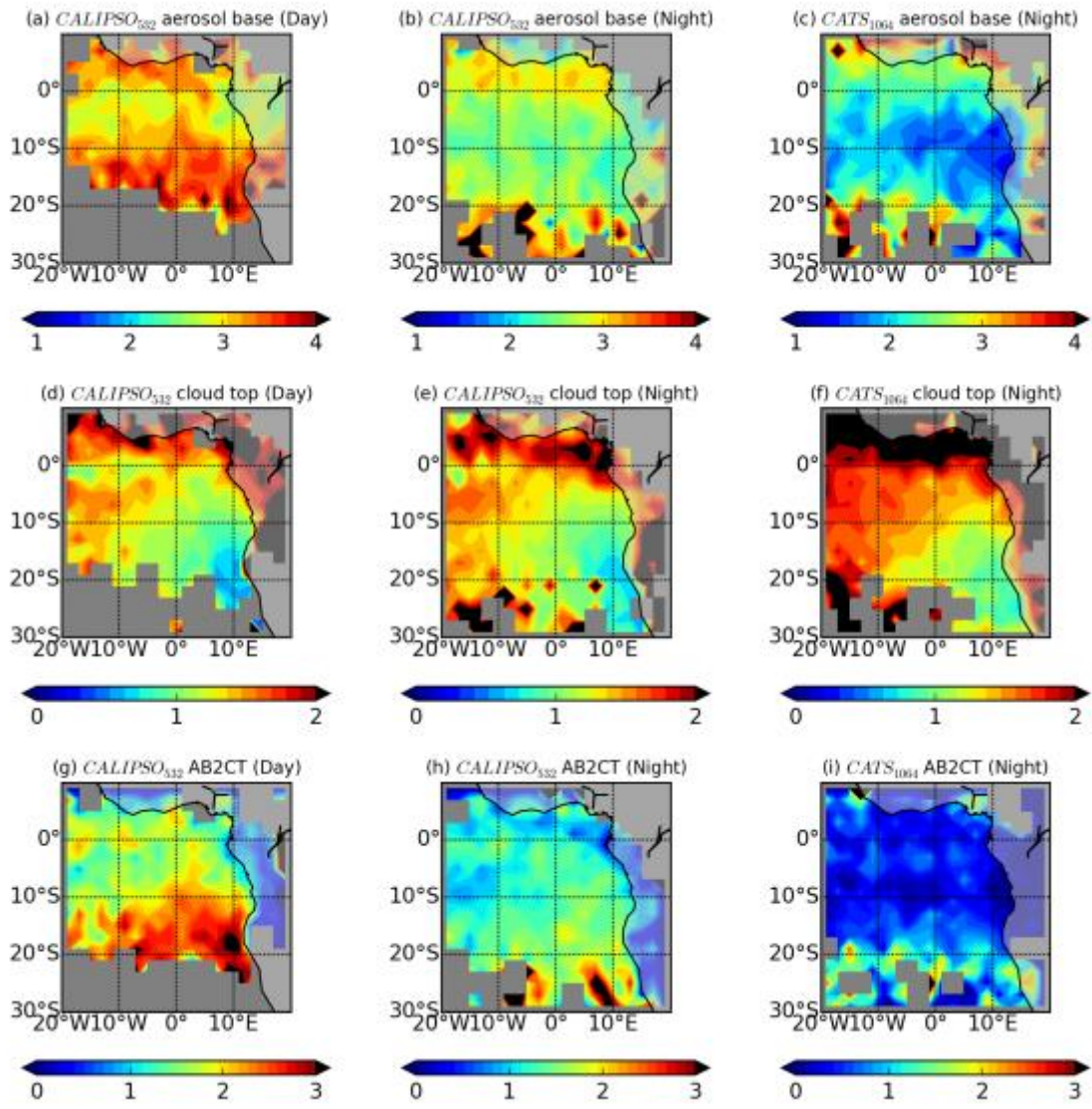
Figure 1 (a) A smoke above MBL cloud event on Aug. 06, 2016. Red dots in the African Continent are fire events. Attenuated total backscatter of CATS 1064nm (b), CALIPSO 532nm (c) and CALIPSO 1064nm. The dashed lines correspond to the point where the CAT and CALIPSO tracks overlap with each other.



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Figure 2 Multi-year (2015-2016) seasonal mean (July to October) cloud fraction (upper row) in the SE Atlantic region based on (a) CALIPSO daytime, (b) CALIPSO nighttime and (c) CATS nighttime observation. The seasonal mean occurrence frequency (lower row) from (d) CALIPSO daytime, (e) CALIPSO nighttime and (f) CATS nighttime observations.

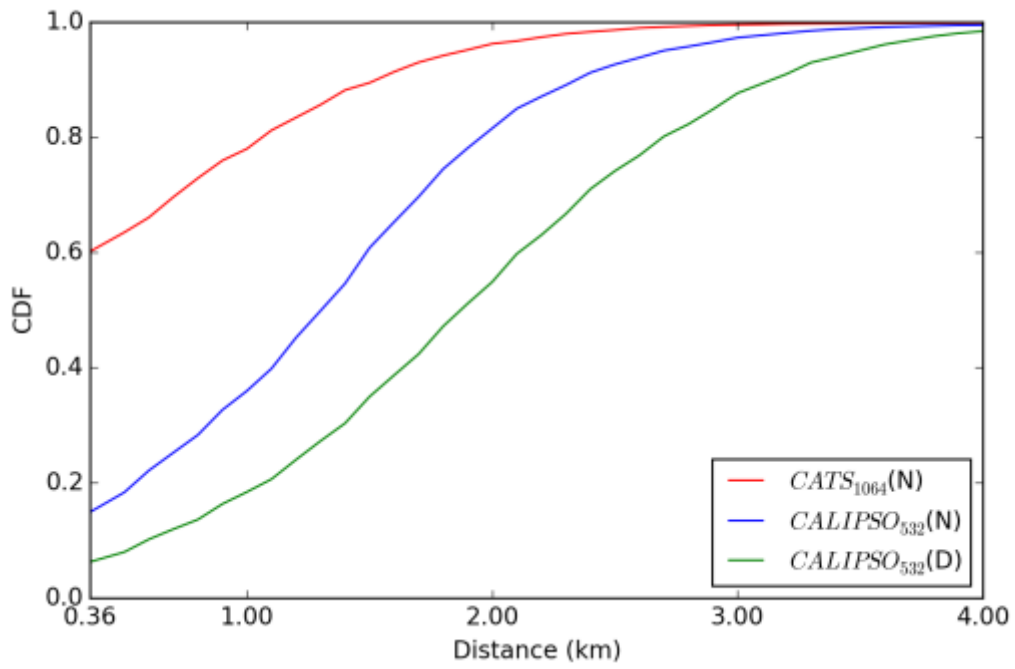
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Figure 3 Multi-year (2015-2016) seasonal mean aerosol layer base height (top row), cloud layer top height (middle row), and aerosol base to cloud top (AB2CT) distance (bottom row) of ACA over the SE Atlantic region during JASO from CALIOP and CATS.

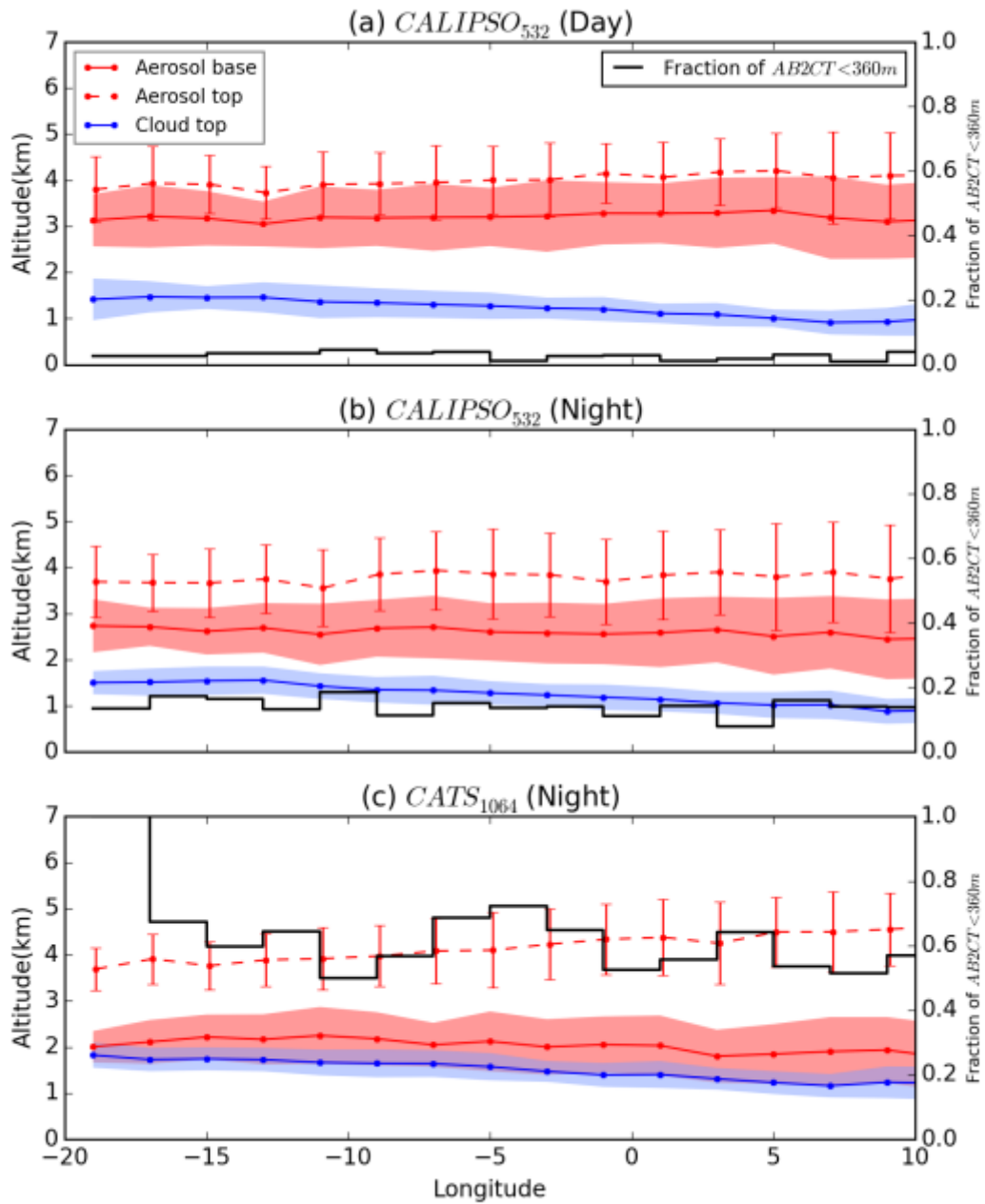
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Figure 4 Cumulative probability distribution function of the distance between aerosol layer bottom and cloud top (AB2CT distance). These curves are derived from the multi-year seasonal ACA data used in Figure 3.

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Figure 5 Meridionally-averaged aerosol bottom (solid red line), top (dashed red line) and cloud top (solid blue line) heights, with fraction of AB2CT<360m (black line), for the SE Atlantic region during JASO, 2015-2016. One standard deviation variability for each are denoted by the red error bars for aerosol top height, and by the red and blue shaded regions for the aerosol bottom and cloud top heights, respectively.

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