

1 Swelling of Transported Smoke from Savanna fires over the Southeast Atlantic Ocean

2 J. Kar^{1,2}, M. Vaughan², J. Tackett^{1,2}, Z. Liu², A. Omar², S. Rodier^{1,2}, C. Trepte², P. Lucker^{1,2}

3 ¹ Science Systems and Applications Inc., Hampton, VA

4 ² NASA Langley Research Center, Hampton, VA

5

6 **Corresponding author:** J. Kar, Science Systems and Applications Inc. (SSAI), NASA LaRC,
7 Hampton, Va 23681, USA. E-mail: jayanta.kar@nasa.gov

8 **Abstract.**

9 We use the recently released Cloud Aerosol Lidar and Infrared Pathfinder Satellite Observations
10 (CALIPSO) Version 4.1 (V4) lidar data to study the smoke plumes transported from Southern
11 African biomass burning areas. Significant improvements in the CALIPSO V4 Level 1 calibration
12 and V4 Level 2 algorithms lead to a better representation of their optical properties, with the
13 aerosol subtype improvements being particularly relevant to smoke over this area. For the first
14 time, we show evidence of smoke particles increasing in size, evidenced in their particulate color
15 ratios, as they are transported over the South Atlantic Ocean from the source regions over Southern
16 Africa. We hypothesize that this is due to hygroscopic swelling of the smoke particles and is
17 reflected in the higher relative humidity in the middle troposphere for profiles with smoke. This
18 finding may have implications for radiative forcing estimates over this area and is also relevant to
19 the ORACLES field mission.

20 **Keywords:** CALIPSO lidar measurements, African smoke, Evolution of optical properties

21

22 **1. Introduction:**

23 The impact of different types of aerosols on our environment is not very well understood and there
24 is an ever-increasing need to characterize the various aerosol types in different parts of the globe
25 (IPCC, 2013). In particular, the smoke from biomass burning needs to be better understood
26 because of the important radiative effects of black carbon (Bond et al., 2013), and because forest
27 fires have been growing in size and frequency in many parts of the world. As such, there have been
28 many studies of biomass burning smoke, their properties and evolution with time (e.g., Reid et al.,
29 2005, Semeniuk et al., 2007, Saide et al., 2015). Some of these properties depend upon the location
30 and type of burning (e.g., smoldering or flaming).

31 Over Southern Africa, savanna burning occurs every year between June and October and
32 constitutes the largest source of biomass burning smoke over the globe (IPCC, 2013; Van der Werf
33 et al., 2010). The smoke plumes from these fires are transported over the Southeast Atlantic Ocean
34 within 5-7 days, overlying one of the largest low altitude extended stratus cloud decks anywhere
35 on the globe, which has consequences for radiative forcing estimates in this area. The direct
36 radiative forcing can be complex under such circumstances changing from cooling in absence of
37 clouds to potentially heating at the top of the atmosphere depending upon the aerosol loading as
38 well as the albedo and fractional coverage of the underlying clouds (Chand et al., 2009, Wilcox,
39 2012, Yu and Zhang, 2013). Passive satellite sensors have generally limited utility here because
40 the aerosol retrievals are done mostly for cloud-free conditions. However, progress has been made
41 in recent years, with researchers exploiting the spectral dependence of the absorption of the
42 upwelling radiation by the aerosols to simultaneously retrieve the cloud and aerosol optical depths
43 for cloudy scenes (Jethva et al., 2013, 2016, Meyer et al., 2015, Sayer et al., 2016). Multi angle
44 polarization information from the POLDER instrument has also been utilized to retrieve the AOD

45 (Waquet et al., 2012, Peers et al. 2015). However, passive sensors can not provide vertically
46 resolved information on these clouds and aerosols, which is crucial for a proper assessment of the
47 radiative forcing, both direct and indirect. This vertical information has become possible in the last
48 decade because of the space borne lidar CALIPSO, which has been providing high quality
49 measurements of the aerosol vertical profiles globally since June 2006 (Winker et al., 2009).
50 Measurements from CALIPSO have been used to derive accurate estimates of radiative forcing
51 of the ACA in this region (Chand et al., 2008, 2009).

52 In the CALIPSO data processing sequence, the attenuated backscatter data are first
53 examined to detect layers using a thresholding algorithm (Vaughan et al., 2009) and then the layers
54 are classified as either a cloud or aerosol (Liu et al., 2009). The aerosol layers are subsequently
55 assigned various subtypes based on their optical properties and geospatial location (Omar et al.,
56 2009). The November 2016 release of Version 4.1 (V4) of the CALIPSO lidar Level 2 data
57 products incorporates significant improvements to the retrieval algorithms, including the aerosol
58 subtype assignments. In particular, there was a significant anomaly in the subtyping over the
59 Southeast Atlantic in earlier versions, where many smoke layers were misclassified as marine
60 layers. This has since been addressed in V4. Many more smoke layers are now identified over the
61 Atlantic, thus presenting a good opportunity for further study of these extensive and regularly
62 occurring smoke plumes. In particular, the evolution of the optical properties of these smoke
63 plumes as they are transported great distances over the South Atlantic may now be better
64 characterized.

65 In this short report, we use V4 CALIPSO data to present evidence of the evolution of size
66 of the smoke particles being exported from the Southern African savanna burning zones. We show
67 that these particles tend to increase in size as they are transported large distances over the ocean.

68 Recent work has noted a close correlation between the smoke plumes and moisture over the South
69 east Atlantic with a general increase in mid-tropospheric moisture in polluted conditions (Adebiyi
70 et al., 2015). While most constituents of smoke plumes are generally hydrophobic, aging and
71 oxidation processes during the transport might make them hydrophilic, and the signatures of this
72 behavior could be discerned in the relative humidity data. This result will have implications for
73 regional radiative forcing as well as for simulations of the transport of these extensive smoke
74 plumes and should be of interest to the currently ongoing ObseRvations of Aerosols above CLouds
75 and their intEractionS (ORACLES) aircraft mission studying the smoke and its interaction with
76 clouds over the same area (Zuidema et al., 2016).

77

78 **2. Data:**

79 We use the CALIPSO V4 level 2 aerosol profile product, which reports height-resolved profiles
80 of the total backscatter and extinction coefficients at 532 nm and 1064 nm, as well as the
81 perpendicular backscatter coefficients at 532 nm for all layers detected. The horizontal resolution
82 of the data is 5km while the vertical resolution is 60m up to 20km and 180m above that. As part
83 of the V4 level 2 updates, the retrieval algorithms were optimized to take maximum advantage of
84 the changes in the V4 level 1 data, released earlier, with significant improvements in both the 532
85 nm and 1064 nm channel calibrations (Getzewich et al., 2015). In particular, the improvement in
86 1064 nm channel calibration makes it feasible to study the optical properties of particles using both
87 the channels with a higher degree of confidence in this new data set. We shall use the particulate
88 color ratio (χ) of the aerosols, which is defined by the relation:

$$89 \quad \chi(z) = \frac{\beta_{1064}(z)}{\beta_{532}(z)} \quad (1)$$

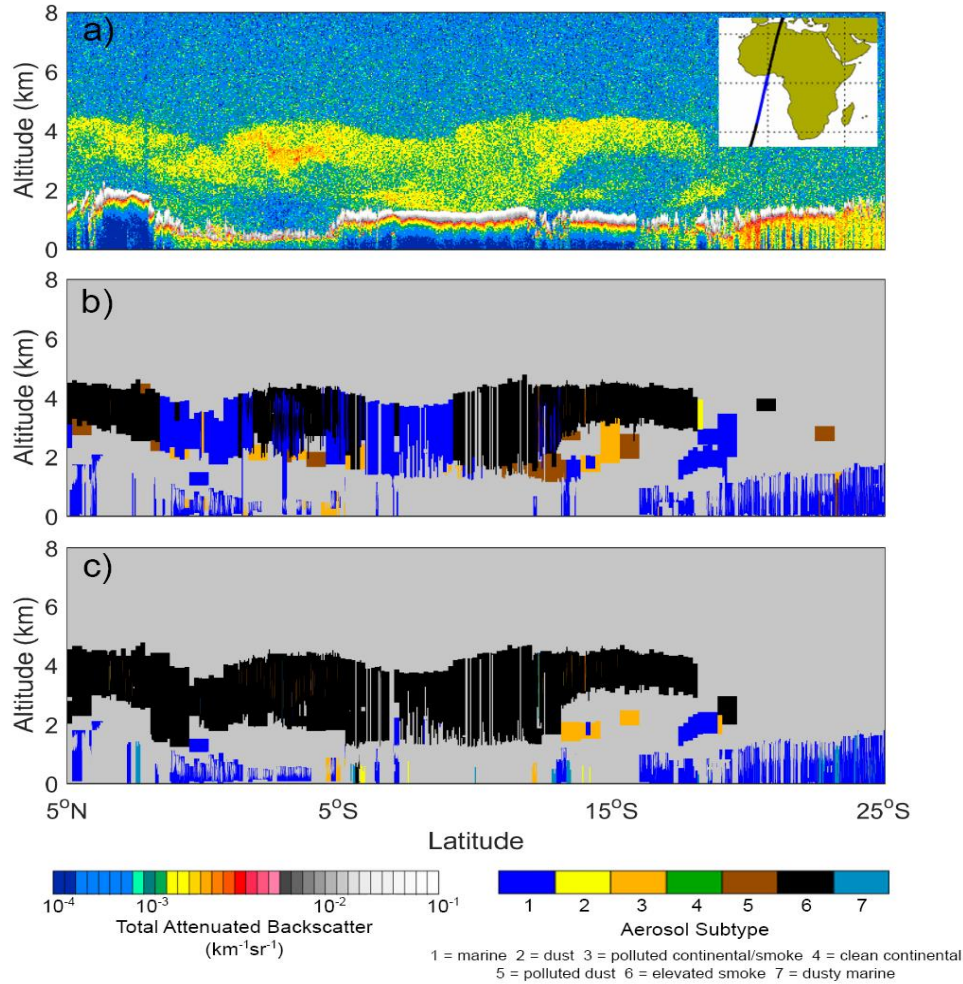
90 where $\beta_{1064}(z)$ and $\beta_{532}(z)$ are the particulate backscatter coefficients retrieved at altitude z at
91 1064 nm and 532 nm respectively. The particulate color ratio values derived from CALIPSO
92 measurements for the various subtypes typically range from 0 to 2, with the small particles like
93 smoke, polluted continental, clean and polluted dust generally having a broad peak near 0.5, while
94 coarse particles like dust, marine and dusty marine have broad peak near 0.7 or so. The
95 uncertainties of the backscatter coefficients are somewhat larger for 1064 nm than 532 nm and the
96 estimated uncertainties for the particulate color ratio for the various subtypes generally peak
97 around 150% - 200%.

98 We use aerosol layers that have been assigned the smoke aerosol subtype by the CALIOP
99 level 2 aerosol classification algorithm for our analysis (Omar et al. 2009).. Previous iterations of
100 the CALIPSO aerosol subtype assignments have been validated by comparison with AERONET
101 data as well as High Spectral Resolution Lidar (HSRL) data (Mielonen et al., 2009, Misra et al.,
102 2013, Burton et al., 2013, Bibi et al., 2016). We also use the 1064 nm measurements retrieved
103 from the Cloud-Aerosol Transport System (CATS) lidar on board the International Space Station
104 (ISS) Mode 7.2 Version 2-00 Level 2 Operational (L2O) Layer and Profile data products. The
105 CATS lidar measures 1064 nm elastic backscatter in polarization planes parallel and perpendicular
106 to the transmitted linearly polarized laser pulses, thus providing depolarization ratio data at 1064
107 nm since March 2015 (Yorks et al., 2016).

108 **3. Results**

109 **3.1. Particulate color ratio evolution in transported smoke.**

110

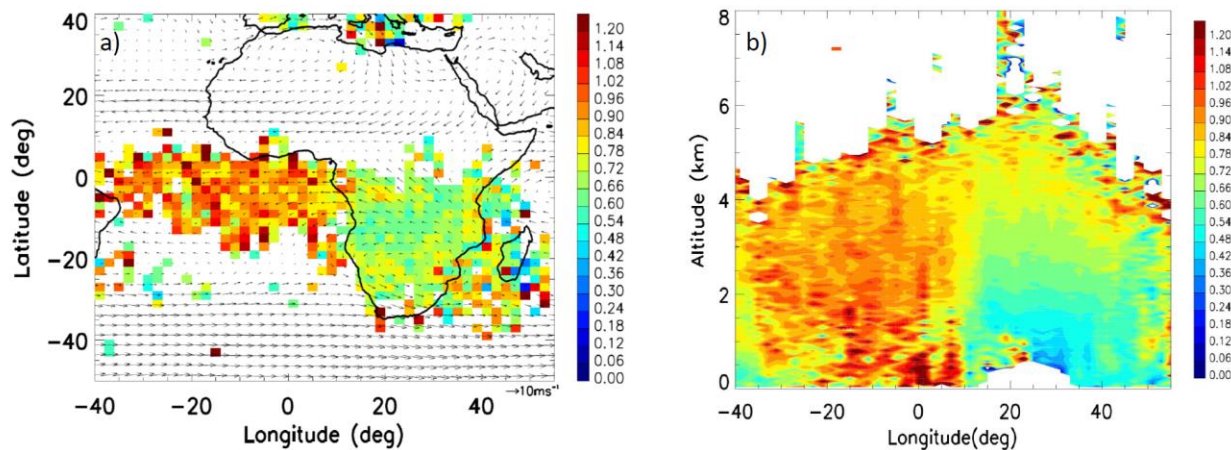


111
 112 **Figure 1.** CALIPSO browse images from September 5, 2010 of a) 532 nm attenuated
 113 backscatter coefficients; and b) aerosol subtypes from V3; and c) aerosol subtypes from
 114 V4.

115 Figure 1a shows the nighttime 532 nm attenuated backscatter coefficients measured over the South
 116 Atlantic Ocean off the coast of Southern Africa on September 5, 2010. The extended plume at 2-
 117 5 km altitude between 19°S and 5°N is smoke that has been transported from the extensive fires
 118 that occur over Southern Africa between June and October every year. Figure 1b shows the aerosol
 119 subtypes assigned in the Version 3 (V3) data products. As can be seen, in the V3 analysis the
 120 plume between 2 and 5 km is punctuated by a large number of misclassified marine layers (in

121 blue). The misclassification of smoke layers as marine was a pervasive problem in V3 data over
122 this area. Figure 1c shows the recently released V4 data, where now we can see a fuller and more
123 coherent smoke plume. The V4 analysis reports a much larger number of smoke layers (and an
124 upward revision of the aerosol optical depth) over this most important and extensive biomass
125 burning area. Thus, we now have more representative information about the spatial extent of
126 biomass burning plumes in this region so that we can better exploit the optical properties reported
127 in the CALIPSO data products.

128



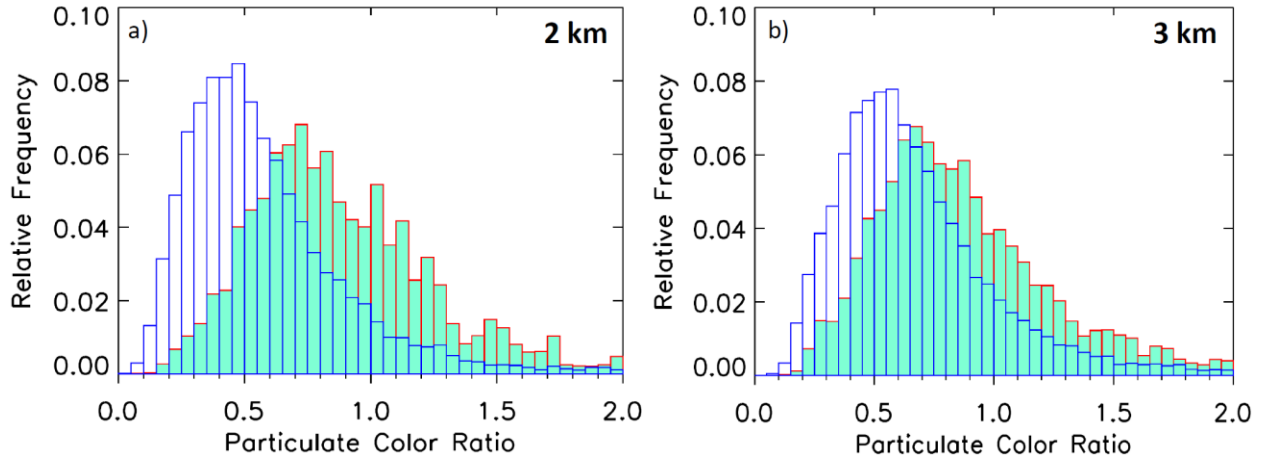
129

130 **Figure 2.** a) Particulate color ratio distribution of smoke at 3 km for August, 2006-2010
131 (binned at 2°x2° in latitude and longitude), with wind vectors at 700 hPa from MERRA-2
132 (August 2006-2010) re-binned into 2.5°x2.5° in latitude and longitude; and b) height-
133 longitude cross section of particulate color ratio along 0-25°S (August 2006-2010). Only
134 nighttime data are used.

135 Figure 2a shows the spatial distribution of the particulate color ratio of the aerosol samples
136 classified as smoke at 3 km using nighttime data for the month of August averaged over 2006-
137 2010. An important change in CALIPSO V4 that is particularly relevant over this area is that the

138 V4 aerosol subtyping algorithm no longer distinguishes between polluted continental and smoke
139 at low altitudes. Instead these layers are identified as “polluted continental/smoke”. An “elevated
140 smoke” subtype is defined for those smoke layers with top altitudes exceeding 2.5 km. We have
141 included both smoke categories in our analysis. As mentioned above, the particulate color ratio is
142 the ratio of the total backscatter coefficients at 1064 nm and 532 nm, and provides a measure of
143 aerosol particle size. The data shown in Figure 2 used only cloud free nighttime profiles. Further,
144 we have included data from only those profiles which had the extinction quality control flag as
145 either zero, indicating that the initial lidar ratio resulted in stable extinction retrievals, or one, which
146 flags those cases where the lidar ratio could be inferred directly from the data (constrained
147 retrievals). We also filtered out the data points where the extinction uncertainty estimate diverged
148 and where the uncertainty of particulate color ratio exceeds 500%. The uncertainty filter retains
149 about 93% of the samples in the region between equator and 35°S and 35°W-55°E. A minimum
150 number of 15 samples was used for each grid box. As can be seen in Figure 2a, there is a clear
151 increase in the particulate color ratio values from the source areas over land to those over the
152 ocean. This likely represents an increase in the size of the smoke particles as they are swept over
153 the ocean over 5-7 days. To our knowledge, this is the first time evidence for such an increase in
154 the size of the smoke particles using satellite data is being reported over this area. The full altitude
155 information can be seen in Figure 2b, which shows the height-longitude cross-section of the
156 particulate color ratios over 0-25°S, using only the cloud free nighttime profiles for August 2006-
157 2010. Once again, the difference between the land and ocean can be clearly seen with somewhat
158 higher values at the lowest altitudes over the ocean, which might be due to gravitational settling
159 of relatively larger and heavier particles.

160



162

163 **Figure 3.** Histograms of particulate color ratio over land (25°S-0,10°E-35°E, in blue) and ocean
164 (25°S-0, 30°W-10°E, in aquamarine, filled) at a) 2 km and b) 3 km (August 2006-2010).

165 Figures 3a and 3b show the histograms of the particulate color ratio at 2 km and 3 km respectively
166 over the source regions on land (in blue, between 25°S-equator, 10°E-35°E) and over oceanic
167 regions (in aquamarine, filled, between 25°S-equator, 30°W-10°E) for August 2006-2010. There
168 is a significant difference in the color ratio distribution between land and ocean with the
169 distribution significantly shifted toward higher values over the ocean at both altitudes with much
170 sharper difference at 2 km. At 3 km, the mean particulate color ratio over land is ~0.7 while that
171 over the ocean is ~0.9, an increase of ~29%, while the increase at 2 km is ~ 60%. This increase in
172 particulate color ratio for smoke particles was seen for all months between June and October and
173 in all years with some interannual variability. Similar results were also obtained using the daytime
174 data although the latter has significantly more noise than the nighttime (not shown). Given that
175 this phenomenon occurs consistently for the key biomass burning months every year, it is not likely
176 to be a data artifact.

177 Apart from the smoke that is transported to the Atlantic, there is another pathway that
178 transports smoke plumes from the southern Africa to the Indian Ocean. This has been noted in
179 satellite data (Swap et al., 1998) and has been known as the “river of smoke”. This outflow can be
180 seen in Figure 2 south of 20°S and between 30°E-55°E. While there is some suggestion of a similar
181 increased color ratio as compared to the land, there is more noise as compared to the outflow to
182 the South east Atlantic. Further, the signature is not clearly seen in the spatial distribution at other
183 altitudes (not shown). Therefore we shall restrict our discussion to the transport to the South East
184 Atlantic ocean only.

185 The current version (V4.1) of the CALIPSO data processing scheme employs the Modern
186 Era Retrospective Analysis for Research and Applications version 2 (MERRA-2) for
187 meteorological information. The latter, for the first time, assimilates aerosol optical depth (AOD)
188 retrieved from AVHRR, MODIS, MISR and AERONET through the integration of the GOCART
189 model and the aerosol radiative feedbacks to the atmospheric fields (Randles et al., 2016). The
190 vertical profiles of the total attenuated backscatter from MERRA-2 generally reproduce the
191 CALIPSO vertical profiles at various places over the globe but show some biases (Randles et al.,
192 2016). Insofar as MERRA-2 already incorporates aerosol information, it is important to determine
193 if the results presented above are biased in any way. We found similar particulate color ratio
194 enhancements over the ocean using V3 CALIPSO data, which reported fewer smoke layers but
195 used GEOS-5.7.2 meteorological data that did not assimilate the aerosol information, thus
196 discounting the possibility of any bias coming from the MERRA-2 meteorology.

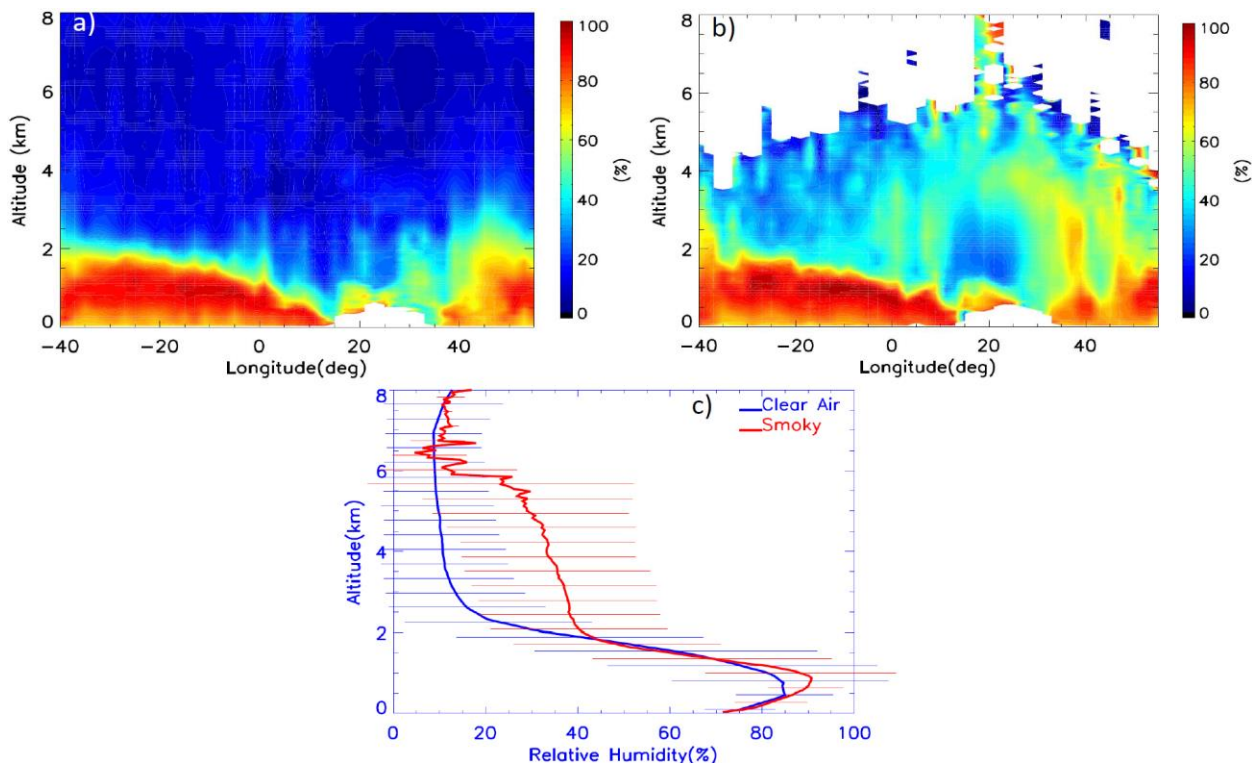
197 The increase in size of smoke particles from CALIPSO observations as seen above is
198 consistent with the findings of Sayer et al. (2014), who studied smoke aerosols transported from
199 biomass burning using data from Analysis of Aerosol Robotic Network (AERONET) stations. For

200 African smoke particles reaching Ascension Island (7.98°S, 14.42°W) in the southeast Atlantic
201 Ocean, they found the volume mean radius ($r_{v,f}$) for fine mode particles to be larger by about 0.02
202 μm than at the inland station of Mongu (15.25°S, 23.15°E) in South Africa. This difference is
203 significantly higher than the estimated uncertainty of 0.01 μm for ($r_{v,f}$) (Sayer et al., 2014). They
204 did not find any evidence of data artifacts resulting from calibration or contamination issues and
205 concluded that the larger radius of the smoke particles at Ascension compared to Mongu is likely
206 to be “a real characteristic of the aerosol transported to this area, rather than an artefact”, with
207 aging over the several days of transport over the Ocean being one of the possible causes.

208 3.2 Relative Humidity Variations

209 The increase in size of the smoke particles over the ocean is likely due to swelling of the particles
210 by water uptake which might have a signature in the relative humidity (RH) profiles.

211



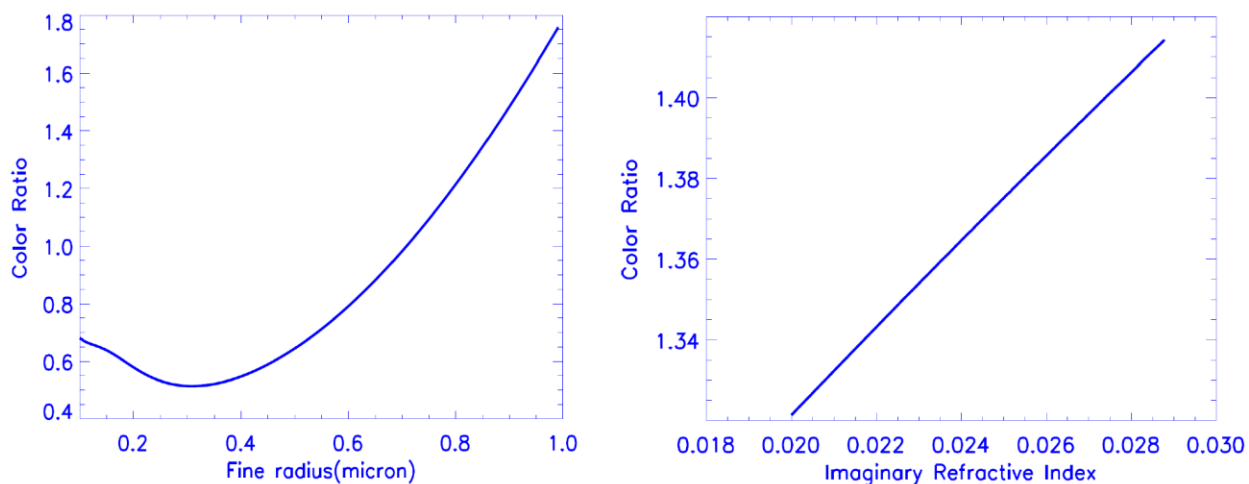
212

213 **Figure 4.** Height longitude cross sections (0-25°S) of relative humidity for a) clear air
214 profiles, b) profiles with smoke samples in them, and c) averaged profiles and standard
215 deviations of relative humidity over the Atlantic Ocean (0-25°S, 30°W-10°E), using all data
216 for August, 2006-2010.

217 Figures 4a and 4b show the height longitude distribution of RH from MERRA-2 as available in
218 CALIPSO data files averaged over 25°S-equator, for August 2006-2010 nighttime data. The clear
219 air RH profiles correspond to cloud free and aerosol free columns within this area, while the smoky
220 profiles correspond to columns that are cloud free but contain smoke samples (essentially
221 corresponding to Figure 2b). Enhanced RH values seem to be associated with the biomass burning
222 smoke plumes. As can be seen in Figure 4c, there is a notable difference between the two mean
223 RH profiles between 2 km and 6 km (over the Atlantic ocean, 0-25°S, 30°W-10°E) where the RH
224 values for the mean smoky profile are larger than in the clear air mean profile. Adebisi et al. (2015)
225 had earlier shown that the RH profiles from MERRA on average tend to reproduce the large scale
226 features from high resolution radiosonde profiles obtained at St. Helena Island (~16°S, 6°W),
227 which is located near the southern parts of this study region. The deviation in the mean RH profiles
228 between MERRA and radiosondes is ~10% (Adebisi et al., 2015). However the bias changes sign
229 around 700 hPa. Below this pressure level, MERRA profiles have a low bias as compared to
230 sondes; above this pressure level, they have a higher bias. Note, however, that Adebisi et al. (2015)
231 used an earlier version of the MERRA product, and not the MERRA-2 reanalyses.

232 The mid-tropospheric difference between the smoky and the clear air RH profiles in Figure
233 4c is quite similar to the results of Adebisi et al. (2015) at St. Helena Island, representing the
234 difference between polluted and non-polluted conditions in September-October. Adebisi et al.
235 (2015) presented individual CALIPSO smoke extinction profiles which often closely matched that

236 of the radiosonde RH profiles at St. Helena with high RH values ($\sim 80\%$) at the top of the smoke
237 layer with the largest extinctions. In contrast, the RH profiles for the non-smoke days showed
238 much lower RH values ($\leq 20\%$) in the mid troposphere. Adebisi et al. (2015) did not discuss the
239 possible swelling effects on the smoke particles, though they did mention the possibility of this
240 occurring. The increase in moisture collocated with increased aerosol loading suggests an
241 environment conducive for swelling of smoke particles. The increase in particulate color ratio of
242 the smoke particles suggesting increased size is thus quite consistent with this scenario.



243
244 **Figure 5.** Change in the color ratio with fine radius and imaginary refractive index from Mie
245 calculations.

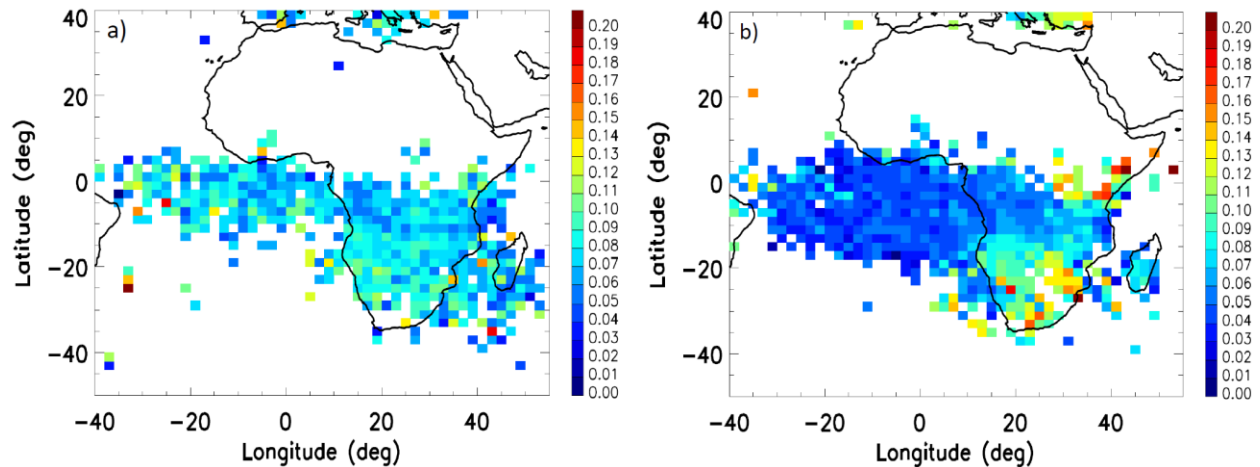
246 In Figure 5a and Figure 5b above we investigate the changes in color ratio as a function of
247 mean fine radius and imaginary refractive index, respectively, using Mie scattering models. The
248 inputs of the fixed parameters, i.e., real part of the refractive index, coarse mode parameters, type
249 and standard deviation of the distributions are from the CALIPSO smoke parameters in Table 1 of
250 Omar et al. [2009]. Figure 5a shows that the color ratio initially decreases until some critical value
251 and then increases rapidly with fine radius. This critical value is quite sensitive to the ratio of fine
252 to coarse mode used in the computation. However this ratio is not well constrained by observations

253 and is highly dependent on the air mass type. Nonetheless, it can be seen that a color ratio
254 increment from about 0.6 to 1.0 indicates a fine radius change of about 0.25 μm or so. Also as the
255 particle takes up water, the imaginary refractive index of the activated particle will decrease since
256 the imaginary refractive index of water is close to zero. Figure 5b shows that the color ratio will
257 tend to decrease with decreasing imaginary refractive index. However, the change in the color ratio
258 as a function of the radius is far larger than as a function of imaginary refractive index. Thus our
259 observation of the increase in color ratio in the moist transported plumes over the Atlantic Ocean
260 is likely to be the result of an increase in the size of the fine mode particles.

261 **3.3. Particulate depolarization of smoke**

262 The particulate depolarization ratio (i.e., the ratio of the backscatter in the perpendicular and
263 parallel channels at 532 nm) reported in the CALIPSO data provides insight into the shape of the
264 scattering particles. The particulate depolarization ratio values for the various aerosol subtypes
265 range between 0 and ~ 0.5 , with a mean value of about 0.04-0.05 for smoke and polluted continental
266 to 0.27 for dust (Liu et al., 2015). Polluted dust has a mean value of about 0.13 and marine near
267 0.02 and dusty marine about 0.15. The values of the estimated uncertainties for the various
268 subtypes peak in the range of 200%-300%. In general, swelling might be expected to enhance the
269 sphericity of particles. However, because biomass burning typically generates quasi-spherical
270 particles having low depolarization ratios (Burton et al., 2013), it may be difficult to detect further
271 changes in particle shape using this measurement.

272



273

274

275 **Figure 6.** a) Particulate depolarization of smoke at 532 nm at 3 km from CALIPSO for
 276 August, 2006-2010 and b) volume depolarization of smoke at 1064 nm at 3 km from CATS
 277 for August 2015-2016. A minimum number of 15 samples per grid box was used for each
 278 plot. Only nighttime were used from both instruments.

279 Figure 6a shows the spatial distribution of the particular depolarization of smoke samples at 3 km
 280 from CALIPSO, once again using only nighttime cloud free profiles in August for 2006-2010. We
 281 rejected depolarization data having estimated relative uncertainty above 500%. This criterion
 282 removes data points with very low negative particulate depolarization with associated uncertainties
 283 much higher than 500%. There is significant noise in the data with only a suggestion of a somewhat
 284 higher depolarization over the land areas over South Africa as compared to the oceanic regions.
 285 As an independent measurement, Figure 6b shows the spatial distribution of the volume
 286 depolarization of smoke at 3 km at 1064 nm as observed by the CATS lidar for August 2015-2016.
 287 The CATS data products do not report particulate depolarization ratios. However, because
 288 molecular contributions to the backscatter signal at 1064 nm are substantially smaller than at 532
 289 nm (by a factor of ~ 17), the CATS 1064 nm volume depolarization ratios should provide

290 essentially the same information as the particulate depolarization ratios. The effect of swelling is
291 seen a bit more clearly in the CATS 1064 nm depolarization ratio data with somewhat higher
292 values over the source regions in southern Africa and falling off over the Atlantic Ocean. The
293 significantly higher depolarization of smoke particles south of 18°S may be related to aging of the
294 particles in the anticyclonic gyre over this region (Figure 2a). In general the burning over Southern
295 Africa takes place in dry conditions and produces soot particles which tend to be non-spherical as
296 compared to the wet burning near the equator (Midzak et al., 2017). Note that CATS data products
297 are only available at 1064 nm, so we cannot confirm the changing color ratio using the CATS data.

298

299 **4. Discussion and Conclusions:**

300 We have presented evidence of an increase in the size of smoke particles that are transported over
301 the South Atlantic Ocean in large amounts from the biomass burning regions of South Africa as
302 reflected in the particulate color ratios retrieved from the CALIPSO space borne lidar. The
303 enhanced RH profiles for smoke samples in the mid troposphere as compared to the clear air
304 samples suggests an association with water uptake by these particles. As such, there have been
305 reports of significantly increased moisture content in biomass burning smoke plumes, particularly
306 for smoldering fires (Achtemeir, 2006, Clements et al., 2006). In Southern Africa, smoldering fires
307 may be more frequent towards the equator during the wet season (Midzak et al., 2017). A number
308 of studies have confirmed the hygroscopicity of smoke under certain conditions. Semeniuk et al.
309 (2007) studied the hygroscopic behavior of 80 aerosol particles sampled from southern African
310 burning sources during the SAFARI 2000 mission, which included tar balls and soot, as well as
311 mixed particles. While tar balls and soot were found to be hydrophobic, mixed particles and
312 particles with inorganic coatings showed significantly enhanced hygroscopicity. A similar

313 conclusion about the effect of inorganic material substantially increasing the hygroscopicity of
314 smoke from Siberian fires was also reached by Popovicheva et al. (2016). Further, Vakkari et al.
315 (2014) found that the hygroscopicity of smoke particles, again sampled from South African
316 biomass burning areas, can increase rapidly within the first 2-4 hours due to oxidation and
317 secondary aerosol formation. Aging and further oxidation of the smoke particles as they are
318 transported to vast distances over the ocean may lead to further water uptake. It is possible that
319 there might be other contributing mechanisms apart from swelling. Radke et al. (1995) had made
320 aircraft observations of a large smoke plume that originated in Oregon and was transported over
321 the Pacific Ocean for about 3 days travelling over 1000 km. They found an increase in the size of
322 the accumulation mode particles and argued that coagulation might have been underestimated as
323 a causative mechanism for size changes in smoke plumes. However Radke et al. (1995) did not
324 consider the possible effects of swelling, which seems to be a more likely explanation for our
325 scenario given the correlation with RH as was also observed by Adebisi et al. (2015) over St.
326 Helena.

327 To our knowledge, this is the first report of a change in size distribution of smoke particles
328 as evidenced from satellite data over this area far from the source regions and is generally
329 consistent with the AERONET finding of a significant increase in $r_{v,f}$ at Ascension Island as
330 compared to Mongu (Sayer et al., 2014). This is an important result, insofar as the aerosol indirect
331 effect depends strongly on the size of the particles (Spracklen et al., 2011). The enhanced moisture
332 associated with the smoke particles may also be important for radiative forcing and leads to a
333 cooling in September-October in this area (Adebisi et al., 2015). The size increase of smoke
334 particles in this area should also provide important constraints for simulations of this southern
335 African smoke transport which show significant discrepancies compared to observations (Das et

336 al., 2017). Therefore, this finding needs to be explored further using field missions as well as with
337 satellite data. In fact, a major field mission, ORACLES, is currently studying the aerosol and
338 cloud properties over this very region, and the ORACLES measurements should provide a wealth
339 of resources to validate the results presented here.

340 **5. Acknowledgements:**

341 The CALIPSO aerosol and cloud profile data as well as the CATS lidar data are available at the
342 NASA Langley Research Center Atmospheric Science Data Center. The MERRA 2 wind data
343 were taken from the MERRA-2 Giovanni instance.

344

345 **References:**

346 Achtemeier, G. L.(2006): Measurements of moisture in smoldering smoke and implications for
347 fog, *Int. J. Wildland Fire*, 15, 517-525.

348
349 Adebisi, A. A. et al. (2015), The convolution of dynamics and moisture with the presence of
350 shortwave absorbing aerosols over the Southeast Atlantic, *J. Clim.*, 28, 1997-2024,
351 doi:10.1175/J-CLI-D-14-00352.1.

352 Bibi, H., K. Alam and S. Bibi, (2016): “In-depth discrimination of aerosol types using multiple
353 clustering techniques over four locations in Indo-Gangetic plains”, *Atmos. Res.*, **181**, 106-
354 114, doi:10.1016/j.atmosres.2016.06.017.

355 Bond, T. C. et al. (2013), Bounding the role of black carbon in the climate system: A scientific
356 assessment, *J. Geophys. Res.*, 118, 5380-5552, doi:10.1002/jgrd.50171.

357 Burton, S. P., et al. (2013), Aerosol Classification from Airborne HSRL and Comparisons with the
358 CALIPSO Vertical Feature Mask. *Atmos. Meas. Tech.*, **6**, 1397-1412.

359 Chand, D. et al. (2008): Quantifying above-cloud aerosol using spaceborne lidar for improved
360 understanding of cloudy-sky direct climate forcing, *J. Geophys. Res.*, **113**, D13206,
361 doi:10.1029/007JD009433.

362 Chand, D., et al. (2009): Satellite-derived direct radiative effect of aerosols dependent on cloud
363 cover, *Nature Geosc.*, **2**, 181-184,doi:10.1038/NGE0437.

364 Clements, C. B., Potter, B. E., and Zhong, S. (2006).: In situ measurements of water vapor, heat,
365 and CO2 fluxes within a prescribed grass fire, *Int. J. Wildland Fire*, **15**, 299-306.

366

367 Das, S., et al., (2017), Biomass burning aerosol transport and vertical distribution over the South
368 African-Atlantic region, *J. Geophys. Res.*, **122**, doi:10.1002/2016JD026421.

369 Getzewich, B. J., et al. (2015): “CALIOP Calibration: Version 4.0 Algorithm Updates”, The 27th
370 International Laser Radar Conference (ILRC 27), *EPJ Web of Conferences*, **119**, 04013,
371 doi:10.1051/epjconf/201611904013.

372 Intergovernmental Panel on Climate Change, IPCC, (2013), *Climate Change 2013: The physical*
373 *science basis. The contribution of working group I to the Fifth Assessment Report of the*
374 *Intergovernmental Panel on Climate Change.*

375 Jethva, H., et al., (2013), A color ratio method for simultaneous retrieval of aerosol and cloud
376 optical thickness of above-cloud absorbing aerosols from passive sensors: Application to
377 MODIS measurements, *IEEE Trans. Geosci. Remote*, **51**, 3862-3870,
378 doi:10.1109/TGRS.2012.2230008.

379 Jethva, H et al., (2016), Validating MODIS above-cloud aerosol optical depth retrieved from
380 “color ratio” algorithm using direct measurements made by NASA’s airborne AATS and
381 4STAR sensors, *Atmos. Meas. Tech.*, 9, 5053-5062.

382 Liu, Z., et al. (2009): The CALIPSO Lidar Cloud and Aerosol Discrimination: Version 2
383 Algorithm and Initial Assessment of Performance, *J. Atmos. Oceanic Technol.*, 26,
384 1198–1213, doi:10.1175/2009JTECHA1229.1.

385 Liu, Z., et al., 2015: Evaluation of CALIOP 532-nm AOD over Opaque Water Clouds, *Atmos.*
386 *Chem. Phys.*, 15, 1265–1288, doi:10.5194/acp-15-1265-2015.

387 Meyer, K., S. Platnick and Z. Zhang, (2015), Simultaneously inferring above-cloud absorbing
388 aerosol optical thickness and underlying liquid phase cloud optical and microphysical
389 properties using MODIS, *J. Geophys Res.*, 120, 5524-5547, doi:10.1002/2015JD023128.

390 Midzak, N. et al. (2017), Determining smoke particle sphericity using CATS data, paper
391 presented at the 97th Annual Meeting of the American Meteorological Society, Seattle,
392 Jan 22-27, 2017.

393 Mielonen, T. et al. (2009). Comparison of CALIOP Level 2 aerosol subtypes to aerosol types
394 derived from AERONET inversion data. *Geophys. Res. Lett.*, 36, L18804.

395 Misra, A. K. et al. (2013). Study of MPLNET-derived aerosol climatology over Kanpur, India
396 and validation of CALIPSO Level 2 version 3 backscatter and extinction products.
397 *J. Atmos. Oceanic Technol.*, **29**, 1285–1294.

398 Omar, A. et al. (2009), The CALIPSO automated aerosol classification and lidar ratio selection
399 algorithm, *J. Atmos. Oceanic Technol.*, 26, 1994-2014.

400

401 Popovicheva, O. B., et al. (2016), Small-scale study of Siberian biomass burning: II. Smoke

402 hygroscopicity *Aerosol and Air Quality Res.*, 16, 1558-1568. Radke, L. F. et al., (1995),
403 Effects of aging on the smoke from a large forest fire, *Atmos. Res*, 38, 315-332.

404 Randles, C. A. et al., (2016), The MERRA-2 aerosol assimilation, NASA/TM-2016-104606 /
405 Vol.45, vol. 45, ed. R. D. Koster.

406 Reid, J. S. et al. (2005), A review of biomass burning emissions part III. Intensive optical properties
407 of biomass burning particles, *Atmos. Chem. Phys.*, 5, 827-849.

408 Saide, P. E. (2015): Central American biomass burning smoke can increase tornado severity in
409 the U.S, *Geophys. Res. Lett.*, **42**, 956–965, doi:10.1002/2014GL062826.

410 Sayer, A. M., et al. (2014), AERONET-based models of smoke-dominated aerosol near source
411 regions and transported over oceans, and implications for satellite retrievals of aerosol
412 optical depth, *Atmos. Chem. Phys.*, 14, 11493-11523.

413 Sayer, A. M. et al., (2016), Extending “deep blue” aerosol retrieval coverage to cases of
414 absorbing aerosols above clouds: Sensitivity analysis and first case studies, *J. Geophys.*
415 *Res.*, 121, 4830-4854, doi:10.1002/2015JD024729.

416 Semeniuk, T. A., et al. (2007), Hygroscopic behavior of aerosol particles from biomass fires using
417 environmental transmission electron microscopy, *J. Atmos. Chem.*, 56, 259-273.

418 Spracklen, D. V. et al., (2011), Global cloud condensation nuclei influenced by carbonaceous
419 combustion aerosol, *Atmos. Chem. Phys.*, 11, 9067-9087.

420 Swap, R. J., et al., (2003), Africa burning: A thematic analysis of the Southern African Regional
421 Science Initiative (SAFARI 2000), *J. Geophys. Res.*, 108, 8465,
422 doi:10.1029/2003JD003747.

423 Vakkari, V., et al. (2014), Rapid changes in biomass burning aerosols by atmospheric oxidation,
424 Geophys. Res. Lett., 41, 2644–2651, doi:10.1002/2014GL059396.

425 Van der Werf, G. R., et al. (2010), Global fire emissions and the contribution of deforestation,
426 savanna, forest, agricultural, and peat fires (1997-2009), Atmos. Chem. Phys., 10, 11707-
427 11735.

428 Vaughan, M., et al. (2009): Fully Automated Detection of Cloud and Aerosol Layers in the
429 CALIPSO Lidar Measurements, J. Atmos. Oceanic Technol., 26, 2034–2050, doi:
430 10.1175/2009JTECHA1228.1.

431 Waquet, F. et al., (2013), Retrieval of aerosol microphysical and optical properties above liquid
432 clouds from POLDER/PARASOL polarization measurements, Atmos. Meas. Tech., 6,
433 991-1016,

434 Wilcox, E. M., (2012), Direct and semi-direct radiative forcing of smoke aerosols above clouds,
435 Atmos. Chem. Phys., 12, 139-149.

436 Winker, D. M. et al., (2009), Overview of the CALIPSO mission and CALIOP data processing
437 algorithms, J. Atmos. Oceanic Technol., 26, 2310-2323.

438 Yorks, J. E. et al. (2016), An overview of the CATS Level 1 processing algorithms and data
439 products, Geophys. Res. Lett., 43, 4632–4639, doi:10.1002/2016GL068006.

440 Yu, H. and Z. Zhang, (2013), New directions: Emerging satellite observations of above-cloud
441 aerosols and direct radiative forcing, Atmos. Environ, 72, 36-40.

442 Zuidema, P. et al., (2016): Smoke and Clouds above the Southeast Atlantic: Upcoming Field
443 Campaigns Probe Absorbing Aerosol’s Impact on Climate, BAMS, 97, 1131–1135,
444 doi:10.1175/BAMS-D-15-00082.1.

445

446

447

448

449

450

451

452

453

454