

1 **Tropical Atlantic Dust and Smoke Aerosol Variabilities related to the Madden-**
2 **Julian Oscillation in MODIS and MISR Observations**

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J. Geophys. Res – Atmosphere

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To be submitted, 04/04/2012

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Abstract

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In this study, MODIS fine mode fraction and MISR non-spherical fraction are used to derive dust and smoke AOT components (τ_{dust} and τ_{smoke}) over the tropical Atlantic, and their variabilities related to the Madden-Julian Oscillation (MJO) are then investigated.

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Both MODIS and MISR show a very similar dust and smoke winter climatology. τ_{dust} is found to be the dominant aerosol component over the tropical Atlantic while τ_{smoke} is significantly smaller than τ_{dust} . The daily MODIS and MISR τ_{dust} are overall highly correlated, with the correlation coefficients typically about 0.7 over the North Atlantic. The consistency between the MODIS and MISR dust and smoke aerosol climatology and daily variations give us confidence to use these two data sets to investigate their relative contributions to the total AOT variation associated with the MJO. However, unlike the MISR dust discrimination, which is based on particle shape retrievals, the smoke discrimination is less certain, based on assumed partitioning of maritime aerosol for both MISR and MODIS.

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The temporal evolution and spatial patterns of the τ_{dust} anomalies associated with the MJO are consistent between MODIS and MISR. The τ_{dust} anomalies are very similar to those of τ anomalies, and are of comparable magnitude. In contrast, the MJO-related τ_{smoke} anomalies are rather small, and the τ_{mar} anomalies are negligible. The consistency between the MODIS and MISR results suggests that dust aerosol is the dominant component on the intra-seasonal time scale over the tropical Atlantic Ocean.

47 **1. Introduction**

48 The Madden-Julian Oscillation (MJO) [*Madden and Julian*, 1971; 1972] is the
49 dominant form of the intra-seasonal (30–90 day) variability in the tropical atmosphere. It
50 is characterized by slow ($\sim 5 \text{ m s}^{-1}$) eastward-propagating, large-scale oscillations in the
51 tropical deep convection over the equatorial Indian Ocean and western Pacific during
52 boreal winter (November–April) [*Lau and Waliser*, 2005; *Zhang*, 2005]. Recently, there
53 is an emerging strong interest in the impacts of the MJO on atmospheric composition
54 [*Tian and Waliser*, 2011], such as aerosol [*Tian et al.*, 2008; 2011], ozone [*Tian et al.*,
55 2007; *Weare*, 2010; *Li et al.*, 2011], carbon dioxide [*Li et al.*, 2010], and carbon
56 monoxide [*Wong and Dessler*, 2007].

57 *Tian et al.* [2008] first examined the aerosol variability related to the MJO using
58 global aerosol products from multiple sensors on various satellite platforms. That study
59 revealed large intra-seasonal variations in the satellite-derived aerosol products over the
60 Tropics, though the interpretation in terms of actual aerosol behavior was ambiguous.
61 *Tian et al.* [2011] further investigated the MJO-related aerosol variability over the
62 tropical Atlantic Ocean using the aerosol optical thickness (AOT) product from the
63 MODIS (Moderate Resolution Imaging Spectroradiometer) on the Aqua satellite. They
64 suggested that the MJO-related intra-seasonal variance accounts for about 25% of the
65 total AOT variance over the tropical Atlantic. They also found that the AOT anomalies
66 are negatively correlated with the low-level zonal wind anomalies over most parts of
67 tropical Atlantic, with the latter leading the former by about 6 days. This indicates the
68 MJO may modulate Atlantic aerosol transport through its influence on the Atlantic low-
69 level zonal winds. Given the potential predictability of the MJO extending to 2–4 weeks

70 [e.g., *Waliser*, 2005], the study by *Tian et al.* [2011] implies that the Atlantic aerosol
71 concentration may be predictable with lead times of 2–4 weeks, which in turn may lend
72 important guidance to predicting air quality, dust storm activity, and ocean nutrient
73 deposition over the Atlantic Ocean.

74 Nevertheless, *Tian et al.* [2011] examined only the total AOT anomalies and did
75 not consider the contribution of different aerosol types to the total AOT anomalies. It is
76 well known that the aerosol over the tropical Atlantic Ocean in the boreal winter season
77 is a mixture of mineral dust from the Sahara desert and the Sahel region, biomass burning
78 smoke from the Sahel and African savanna regions, and marine aerosol (primarily sea salt
79 and secondly sulfate aerosols) from the ocean surface [*Kaufman et al.*, 2002; 2005a;
80 2005b]. Since dust, biomass burning smoke and marine aerosols play very different roles
81 in the radiative forcing and cloud formation process, it is of great interest to partition the
82 total aerosol into individual aerosol components and examine the MJO-related variability
83 in each aerosol type.

84 Previous studies [*Kaufman et al.*, 2002; 2005a; 2005b] have suggested that
85 satellite data distinguishing fine-mode aerosols from coarse-mode aerosols could be used
86 to separate the aerosol into specific types, since different aerosol types (e.g., smoke, dust,
87 and maritime sea salt aerosols) have different fine mode fraction (FMF) values. FMF is
88 the fraction of total AOT contributed by the fine-mode aerosols. For example, smoke
89 from wildfire biomass burning and urban-industrial air pollution from human combustion
90 processes are dominated by fine-mode particles. On the other hand, desert dust and sea
91 salt, mainly a result of wind and erosion processes, have a much larger proportion of
92 coarse-mode particles. *Kaufman et al.* [2002; 2005a; 2005b] first demonstrated the

93 usefulness of combining the total AOT and FMF measurements from MODIS/Terra to
94 estimate dust and biomass burning smoke AOT components (see detailed description in
95 section 2). This method has since been widely adopted by the community to understand
96 aerosol types and their climate forcing [Bellouin *et al.*, 2005, among many others].

97 Measuring the reflectance of a target from different directions is very useful
98 because geophysical media, for example aerosols, reflect solar light differently in
99 different directions. The variations between the reflectances acquired from a variety of
100 observation angles can be interpreted (with appropriate models) in terms of aerosol
101 properties such as particle size, shape, and single-scattering albedo [Kahn *et al.*, 1998,
102 2001; Chen *et al.*, 2008]. In particular, MISR's sensitivity to the characteristics of the
103 aerosol scattering phase functions enables it to distinguish between the non-spherical and
104 spherical particles, and thus provides a possible way to separate mineral dust aerosols
105 from other aerosol components. A series of studies has explored the ability of MISR to
106 retrieve mineral dust properties theoretically [Kahn *et al.*, 1997, 2001; Kalashnikova *et*
107 *al.*, 2005; Kalashnikova and Kahn, 2006], the sensitivity of the theoretical results, as well
108 as the application of non-spherical dust models for five Saharan dust field events over the
109 Atlantic Ocean [Kalashnikova and Kahn, 2008].

110 In this study, we use MODIS FMF to separate the total column AOT into its
111 individual types over the tropical Atlantic, especially the dust and biomass burning
112 smoke aerosols, because of its more frequent sampling. We will also use MISR aerosol
113 non-spherical fraction to derive the dust aerosol component because of its greater particle
114 type information content, and compare it with that derived from MODIS. The resulting
115 intra-seasonal variabilities associated with the MJO in dust and biomass burning smoke

116 aerosols over the tropical Atlantic Ocean are then examined.

117 The rest of this paper is organized as follows. Section 2 describes the MODIS and
118 MISR data, and the methodology used to derive specific aerosol components from the
119 total AOT. The climatology of specific aerosol component as well as the comparison of
120 dust aerosol between MODIS and MISR is presented in Section 3 in order to examine the
121 fidelity of the methods described in Section 2. The main results of this paper, the MJO-
122 related dust and smoke aerosol anomalies, are presented in Section 4. Conclusions and
123 discussions are presented in Section 5.

124 **2. Methodology and Data Description**

125 **2.1. MODIS**

126 *2.1.1. Method to Calculate Dust, Smoke AOT*

127 The method developed by *Kaufman et al.* (2005a, b) to derive the AOTs for dust
128 and biomass burning smoke using MODIS total AOT and FMF measurements is briefly
129 described as below. Over the tropical Atlantic Ocean during the boreal winter season, the
130 aerosol is a mixture of dust, biomass burning smoke, and marine aerosols. Note that the
131 term ‘anthropogenic aerosol’ in *Kaufman et al.*’s papers is replaced by ‘biomass burning
132 smoke’ as the latter is far more appropriate term for the region of study. However, the
133 ‘smoke’ does include a small contribution from air pollution that originates primarily
134 from the US and South American continents. With the two constraints that both the total
135 AOT and its fine model fraction can be partitioned into contributions from the dust,
136 smoke, and marine aerosol components, we have the following equations:

$$137 \quad \tau = \tau_{\text{dust}} + \tau_{\text{smoke}} + \tau_{\text{mar}} \quad (1)$$

$$138 \quad f \times \tau = f_{\text{dust}} \times \tau_{\text{dust}} + f_{\text{smoke}} \times \tau_{\text{smoke}} + f_{\text{mar}} \times \tau_{\text{mar}} \quad (2),$$

139 in which τ and f denote AOT and FMF respectively (τ and AOT as well as f and FMF are
 140 interchangeable in this paper), and the subscripts ‘dust’, ‘smoke’, and ‘mar’ indicate dust,
 141 smoke, and marine aerosol component, respectively. Rewriting (1) and (2), we get

$$142 \quad \tau_{\text{dust}} = [\tau \times (f_{\text{smoke}} - f) - \tau_{\text{mar}} \times (f_{\text{smoke}} - f_{\text{mar}})] / (f_{\text{smoke}} - f_{\text{dust}}) \quad (3)$$

$$143 \quad \tau_{\text{smoke}} = [\tau \times (f - f_{\text{dust}}) - \tau_{\text{mar}} \times (f_{\text{mar}} - f_{\text{dust}})] / (f_{\text{smoke}} - f_{\text{dust}}) \quad (4).$$

144 With τ and f being MODIS measurements, τ_{dust} and τ_{smoke} can be computed directly if f_{dust} ,
 145 f_{smoke} , f_{mar} , and τ_{mar} are known.

146 In *Kaufman et al.*’s study, the FMFs of dust, smoke, and marine aerosols are
 147 assumed to be constant, and were derived by averaging the MODIS/Terra FMF
 148 measurements over selected regions and time periods where one specific aerosol type
 149 dominates, with their uncertainties estimated from these selected measurements:
 150 $f_{\text{dust}}=0.5\pm0.05$, $f_{\text{smoke}}=0.9\pm0.05$, and $f_{\text{mar}}=0.3\pm0.1$ [2005b]. However, the actual FMF
 151 values vary with season and location [*Maring et al.*, 2003; *Jones and Christopher*, 2007,
 152 2011; *Yu et al.*, 2009], thus large uncertainties are expected from using constant f_{dust} ,
 153 f_{smoke} , f_{mar} when applying Kaufman’s formula. More discussion on this issue as well as
 154 the sensitivity of our results to the FMF values will be given in Section 4.3.

155 The marine AOT, τ_{mar} , depends strongly on surface wind speed as its primary
 156 component is sea-spray salt [e.g., *Smirnov et al.*, 2003]; it is estimated using the empirical
 157 formula in *Kaufman et al.* [2005b]:

$$158 \quad \tau_{\text{mar}} = 0.007W + 0.02 \quad (5)$$

159 Here, W is the surface (10 meter) wind speed from the ECMWF ERA-Interim reanalysis
 160 [*Dee et al.*, 2011]. The global mean of τ_{mar} is around 0.06 ± 0.005 [e.g., *Kaufman et al.*,

161 2001].

162 2.1.2. MODIS Data Description

163 MODIS instruments are operating on both the Terra and Aqua satellites. Each has
164 a viewing swath width of 2,330 km and views the entire surface of the Earth every one to
165 two days. In this study, daily AOT and FMF measurements at 0.55 μm from the
166 MODIS/Aqua Level-3 Collection 5.1 dataset [Levy *et al.*, 2009] on $1^\circ \times 1^\circ$ spatial grids
167 are used. The uncertainties of τ are $\pm (0.03 + 0.05\tau)$ over ocean and $\pm (0.05 + 0.15\tau)$ over
168 land. In MODIS, the fine-mode aerosols refer to the aerosols with a size distribution of
169 radii centered between 0.1 and 0.25 μm , whereas the coarse-mode aerosols refer to those
170 with a size distribution of radii centered between 1 and 2.5 μm . Validation studies
171 indicate that, over ocean, the uncertainties of f are large for low AOT ($\tau < 0.15$) but
172 typically less than about 20% for large AOT ($\tau > 0.15$) [Kleidman *et al.*, 2005; Remer *et*
173 *al.*, 2005]. Over land, MODIS does not provide any quantitative information about the
174 aerosol size [Levy *et al.*, 2010]. As a result, we use the τ and f data over the ocean only.
175 The period of 4 July 2002 to 1 June 2009 is used for consistency with the study by Tian
176 *et al.* [2011].

177 Note that we use MODIS/Aqua aerosol data rather than MODIS/Terra data even
178 though MISR is on board Terra. One reason is that we want to be consistent with the
179 study in Tian *et al.* [2011]. More importantly, MODIS/Terra and MISR aerosol retrievals
180 are much less spatially overlapped due to the exclusion of sun glint regions over the
181 ocean in MODIS. As a result there are many fewer days on which both MODIS/Terra and
182 MISR made observations of a given region, compared to the situation between
183 MODIS/Aqua and MISR. Since the direct comparison between daily MODIS and MISR

184 aerosol data serves as an essential part of this study, we chose to use MODIS/Aqua data.
185 The results based on MODIS/Terra were compared with those based on MODIS/Aqua,
186 and it is found that the difference is negligible for our application.

187 2.1.3. MODIS Data Rejection

188 In this work, some MODIS τ and f measurements are rejected in our calculations
189 for the reasons described below.

190 The dependence of τ_{dust} and τ_{smoke} on τ and f based on equations (3) and (4) is
191 illustrated in Figure 1, where τ_{mar} is set to 0.06 (the approximate global mean value of
192 τ_{mar}). It is seen that the larger the τ , the larger τ_{dust} and τ_{smoke} . Also, as f increases, τ_{smoke}
193 increases and τ_{dust} decreases. Figure 1 also suggests that equations (3) and (4) can produce
194 reasonable (nonnegative) values for both τ_{dust} and τ_{smoke} only when the paired τ and f
195 measurements fall within a limited region bounded by the white dashed lines ($\tau_{\text{dust}} = -0.03$
196 and $\tau_{\text{smoke}} = -0.03$). We relax the limit of τ_{dust} or τ_{smoke} to -0.03 instead of 0 because the
197 uncertainties of the MODIS AOT are ± 0.03 over ocean, thus a small negative AOT up to
198 -0.03 is regarded as indistinguishable from the value 0 [Remer *et al.*, 2005]. Note that
199 calculated τ_{dust} and τ_{smoke} in the range of 0 to -0.03 have been set to zero. Outside this
200 region, either τ_{dust} or τ_{smoke} is too negative, and correspondingly, τ_{smoke} or τ_{dust} would be
201 larger than the total AOT, which is not physical. Therefore, the τ and f measurements
202 giving rise to such τ_{dust} and τ_{smoke} values are rejected. In addition, τ measurements greater
203 than 2 are excluded, given the large uncertainties from cloud contamination [Zhang *et al.*,
204 2005].

205 The count distribution of paired MODIS τ and f measurements as a function of τ
206 and f is summarized for the tropical Atlantic ($20^{\circ}\text{S} - 30^{\circ}\text{N}$, $60^{\circ}\text{W} - 20^{\circ}\text{E}$) and for the

207 2002–2009 boreal winters (October to April) in Figure 2a. It is found that the majority of
208 observations fall within the region bounded by two lines ($\tau_{\text{dust}} = -0.03$ and $\tau_{\text{smoke}} = -0.03$);
209 however, there is also a large number of observations falling outside the two lines, which
210 are rejected in the calculation. The rejection results from both the limitations in MODIS
211 data and the limitations in Kaufman’s method. For example, in some cases τ is typically
212 between 0.1–0.3, which is much larger than the baseline τ_{mar} , while f is extremely small,
213 nearly 0 (see bottom of Figure 2a). These observations are very likely artifacts of cloud
214 contamination [e.g., *Zhang et al.*, 2005; *Tian et al.*, 2008]. For other cases, τ is also
215 significantly larger than τ_{mar} and f falls between f_{mar} and f_{dust} , suggesting the aerosol is a
216 mixture of dust and sea salt (see the area between the line of $\tau_{\text{smoke}} = -0.03$ and the line of
217 f_{mar}). These are likely valid observations; however, due to the difficulties in separating
218 dust from sea salt over this regime [*Kaufman et al.* 2005b], we don’t include these
219 observations in our calculations.

220 The percentage of days with MODIS τ and f measurements available relative to
221 the total days during 2002–2009 boreal winters (1269) is shown in Figure 2b. The
222 occurrence of cloudy conditions affecting the retrieval of MODIS aerosol parameters is
223 evident, with the fraction of observations decreasing from more than 70% over the clear
224 subtropical North Atlantic to about 35% over the cloudy equatorial Atlantic. After
225 applying the data rejection described above, overall more than 50% of the observations
226 remain. The most frequent rejections occur over the subtropical oceans, possibly due to
227 the ambiguity between sea salt and dust aerosol over those regions

228 It is noted that we applied *Kaufman et al.*’s formula in a stricter way than what
229 was originally done by *Kaufman et al.* [2005a and 2005b]. In *Kaufman et al.* [2005a]

230 when computing τ_{smoke} , they set negative τ_{smoke} values to zero, regardless of their
231 magnitude. And in *Kaufman et al.* [2005b], when computing τ_{dust} , f is forced to be
232 bounded within f_{dust} and f_{smoke} by setting any f values outside the bounds to the limiting
233 values, to guarantee τ_{dust} is always positive. In our work, we include only those
234 observations that fit the formula well. Nevertheless, it should be noted that although the
235 data rejection we applied removes a large number of suspicious observations, doing so
236 does not in itself guarantee the accuracy of the remaining applications, given that f_{dust} ,
237 f_{smoke} , and f_{mar} are assumed constant, and τ_{mar} was estimated from an empirical formula.
238 For the former factor, sensitivity tests will be performed to examine the extent to which
239 our results are sensitive to the variations in FMF values. For the latter, the deviations of
240 the actual τ_{mar} (mainly sea salt) compared to that empirically computed will be interpreted
241 primarily as changes in dust amount, thus affecting the accuracy of τ_{dust} . However, as will
242 be presented next, MISR can distinguish dust from sea salt based on retrieved particle
243 shape, thus τ_{dust} derived from MISR is not subject to the ambiguity between dust and sea
244 salt that affects Kaufman's method. In this sense the examination of MISR aerosol
245 observations provides validation for the more extensive MODIS data set, in addition to
246 offering actual results independent of MODIS.

247 **2.2. MISR**

248 The MISR instrument aboard the Terra satellite views Earth in four spectral bands
249 in each of nine view angles spread out in the forward and aft directions along the flight
250 path at 70.5°, 60.0°, 45.6°, 26.1°, and nadir. With a swath about 360 km wide, global
251 coverage with MISR is acquired about once every 9 days at the equator and 2 days
252 toward the poles. Data collected from all nine viewing angles provide independent

253 constraints on aerosol properties, separating mineral dust aerosols from other aerosol
254 components operationally [*Kahn et al.*, 2010].

255 Assuming the mineral dust is all non-spherical and the non-spherical part of AOT
256 is all from dust, MISR τ_{dust} can be directly computed as the non-spherical fraction of the
257 total AOT:

$$258 \quad \tau_{\text{dust}} = \tau \times f_{\text{non-spherical}} \quad (6).$$

259 In MISR, smoke and marine aerosols (both sea salt and sulfate aerosols)
260 contribute to the spherical part of the total AOT. (Refer to Table 3 in *Kahn et al.* [2001]
261 for more details on shape categories of different aerosol components.) If τ_{mar} is again
262 taken as computed from equation (5), τ_{smoke} can be approximated as the difference
263 between the total τ and τ_{dust} and τ_{mar} :

$$264 \quad \tau_{\text{smoke}} = \tau - \tau_{\text{dust}} - \tau_{\text{mar}} \quad (7).$$

265 However, it should be kept in mind that unlike MISR τ_{dust} , the MISR τ_{smoke} derived this
266 way, and all the MODIS particle type distinctions, are not independent observations
267 based on actual physical constraints, and their accuracy is subject to the large
268 uncertainties due to the empirically calculated τ_{mar} .

269 We use MISR-derived Level 3 daily τ and non-spherical fraction at 0.558 μm on
270 $0.5^\circ \times 0.5^\circ$ spatial grids during the same time period as the MODIS data. MISR τ and τ_{dust}
271 are interpolated onto $1^\circ \times 1^\circ$ grids to compare with MODIS results. A number of
272 validation studies have shown that overall, about 70% to 75% of MISR AOT retrievals
273 fall within 0.05 or 20% \times AOT, and about 50% to 55% are within 0.03 or 10% \times AOT,
274 except at sites where mixed dust and smoke are commonly found [*Kahn et al.*, 2010].

275 Particle property validation suggests that expected MISR sensitivity to the spherical
276 versus non-spherical particles is about 20% for AOT above 0.15, and diminishes for mid-
277 visible AOT below this value [*Kahn et al.*, 1997; *Kalashnikova et al.*, 2005;
278 *Kalashnikova and Kahn*, 2006]. Thus we only use $\tau \geq 0.15$ to calculate τ_{dust} from equation
279 (6). With this τ cutoff, it is found that more than 70% of the observations still remain for
280 calculation over most of the tropical Atlantic (Figure 3b). Note generally MISR τ
281 retrievals are available for about 15% of the time, except over the convectively active
282 regions (decreased to about 10%, Figure 3a). Note also that although it is never been
283 explicitly addressed, MODIS sensitivity to particle properties, e.g. FMF, also diminishes
284 at low AOT (implied in Figure 7 in *Kahn et al.* [2009]), and MISR actually has greater
285 sensitivity at low AOT than MODIS due to the long atmospheric paths observed by its
286 steeper-viewing cameras.

287 Note that MISR also retrieves aerosol particle size information, thus conceptually
288 it is possible to follow the same method utilized for MODIS to separate the total AOT
289 into specific types in MISR. However MISR categorizes the aerosol particles into three
290 bins: “fine” (particle radii < 0.35 μm), “medium” (radii between 0.35 and 0.7), and “large”
291 (radii > 0.7 μm) modes, instead of two bins as “fine” vs. “coarse” in MODIS. Thus
292 applying Kaufman’s formula to MISR would require considerable additional work, but
293 not necessarily lead to greater insight, because similar assumptions would be required to
294 apply the size-discrimination method to MISR as to MODIS. Furthermore, the different
295 radii range for MODIS and MISR “fine” mode would make it impossible to cross-
296 validate the FMF values between these two datasets. Therefore, using MISR aerosol size
297 information to separate different aerosol components following Kaufman’s method is

298 considered beyond the scope of current paper.

299 **3. Comparison of Aerosol Components between MODIS and MISR**

300 In this section, the AOTs of individual aerosol components over the tropical
301 Atlantic Ocean are examined, and the MODIS and MISR results are compared.
302 Climatological maps are examined first, in order to investigate how well the methods
303 presented in Section 2 capture the basic features of the long-term mean.

304 **3.1. Aerosol Winter Climatology**

305 *3.1.1. MODIS*

306 The MODIS climatological mean (2002–2009) boreal winter (November to April)
307 aerosol maps over the tropical Atlantic Ocean for τ , τ_{dust} , τ_{smoke} , and τ_{mar} , as well as their
308 percentages relative to τ , are shown in Figure 4. The climatological mean τ features a
309 zonally oriented, optically thick aerosol plume centered at around 5°N–8°N stretching
310 across the equatorial Atlantic Ocean. The magnitude and latitudinal extent are greatest
311 over the eastern equatorial Atlantic along the west coast of Africa and gradually decrease
312 westward toward the central and western equatorial Atlantic, as expected for an aerosol
313 plume that originates in Africa (Figure 4a). Note that this map is similar to Figure 1b
314 (color shaded area) in *Tian et al.* [2011], but is not identical, due to the data rejection
315 performed in this study. The magnitude of τ in Figure 4a is about 80% of that in *Tian et*
316 *al.* [2011] over the heavy aerosol loading region ($\tau > 0.15$), and comparable over the clean
317 subtropical Atlantic. Nevertheless, their overall patterns are similar, with spatial
318 correlation as high as 0.98. The spatial pattern of climatological τ_{dust} resembles the τ
319 pattern greatly, with maximum τ_{dust} found along the west coast of Africa (about 0.35) and
320 a gradual decrease toward the central and western equatorial Atlantic (Figure 4b). Over

321 the equatorial Atlantic Ocean, where aerosol loading is high ($\tau > 0.15$), dust is the
322 dominant aerosol component, contributing greater than 50%, and as much as 75%, to τ
323 (Figure 4c).

324 Compared to τ_{dust} , τ_{smoke} is significantly weaker. Plumes of fine-mode-dominant
325 aerosol are found originating from the African (biomass burning smoke) as well as the
326 South American continents (air pollution) (Figure 4d). A contribution of more than 20%
327 to total τ is found over the eastern tropical Atlantic, while less than 20% over the western
328 part (Figure 4e). Figure 4f shows that τ_{mar} is very small (about 0.04) over the Atlantic
329 intertropical convergence zone and the west coast of Africa because of the weak trade
330 winds. Over the clean subtropical Atlantic, marine aerosol is the dominant component
331 (>50%) due to lack of dust and smoke aerosols over these regions.

332 These results indicate that the major aerosol plume over the equatorial Atlantic
333 Ocean in Figure 4a is the dust originating in the Sahara desert and the Sahel region, with
334 some contribution of biomass burning smoke originating in the Sahel and African
335 savanna. These aerosol distributions are generally consistent with previous observational
336 results [e.g., *Husar et al.*, 1997; *Kaufman et al.*, 2005b; *Huang et al.*, 2010a], lending
337 some confidence to the quantitative assessment of the relative dust, smoke and marine
338 AOTs contributions to the total AOT from equations (3) and (4) from MODIS.

339 3.1.2. MISR

340 MISR winter climatologies of τ , τ_{dust} , and τ_{smoke} are shown in Figure 5. Recall that
341 for these results only cases with $\tau \geq 0.15$ are used due to the fact that sensitivity to shape
342 diminishes when aerosol concentration is low. Data rejection is also applied to MODIS τ ,
343 too, as discussed in Section 2.1.3. Thus, we don't expect Figure 5a to be the same as

344 Figure 4a. Nevertheless, we do find they have highly consistent spatial patterns and
345 comparable magnitudes. The climatological MISR τ_{dust} map (Figure 5b) is also very
346 similar to the MODIS τ_{dust} (Figure 4b). The MISR τ and τ_{dust} are slightly larger than the
347 MODIS counterparts partly due to MISR's exclusion of low aerosol cases, but more
348 important reasons will be addressed in Section 3.2. It is seen that MISR τ_{dust} contributes
349 more than 40%, up to more than 60%, to the total τ for cases with $\tau \geq 0.15$, further
350 confirming that dust is the dominant aerosol component over the equatorial Atlantic
351 Ocean (Figure 5c). It is also noted that MISR has 10% to 15% lower dust fraction over
352 the equatorial Atlantic. Again, this difference could be partly attributed to the τ cutoff in
353 MISR, but more importantly could result from both uncertainties involved in the
354 derivation of MODIS and MISR τ_{dust} , especially in MODIS, given the assumptions used.
355 This point will be discussed again in Section 3.2. The climatological MISR τ_{smoke} ,
356 calculated as described in Section 2.2, has a 10% larger contribution to total τ than
357 MODIS τ_{smoke} over the equatorial Atlantic (Figure 5d and 5e compared to Figure 4d and
358 4e).

359 The overall consistency between the MODIS and MISR aerosol climatology gives
360 us confidence in applying two independent satellite datasets and methods to derive the
361 dust and smoke aerosols over the tropical Atlantic Ocean. However, the consistency in
362 climatology does not guarantee their consistency on shorter time scales, for example, on
363 the intra-seasonal time scale of importance here. Therefore, in next subsection we directly
364 compare the coincident daily MODIS and MISR τ_{dust} over the tropical Atlantic Ocean.

365 **3.2. Comparison of Coincident MODIS and MISR Dust Aerosols**

366 In order to compare the daily τ_{dust} derived from MODIS and MISR, the

367 correlation between them for all their coincident days during 2002-2009 over the tropical
368 Atlantic is calculated. Here, the term “coincident” simply means the MODIS and MISR
369 observations fall within a same grid box and on a same day, which is less strict than its
370 usual implication used by the satellite community. We use the whole years for 2002-2009
371 instead of winters only in order to obtain as many coincident days as possible. The
372 correlation is only calculated when there are at least five coincident days at a grid point.

373 Figure 6a shows that overall the MODIS and MISR dust AOTs are well
374 correlated. The correlation is systematically higher in the north Atlantic region than that
375 in the south, with more coincident days over the north Atlantic (Figure 6b). Over the
376 North Atlantic, the correlation coefficient is typically around 0.7 or larger, whereas over
377 the South Atlantic, the correlation is typically less than 0.5 with quite a few spots less
378 than 0.2, or even negative. Generally, the correlation is higher when it is closer to the
379 African continent, and decreases gradually as the dust is transported away from the
380 source of the dust load. These results suggest that the MODIS and MISR derived dust
381 aerosols agree with each other quite well over heavy dust load regions, whereas they are
382 less consistent over the regions with less frequent dust occurrence or small dust aerosol
383 concentration.

384 Figure 6a shows that overall the MODIS and MISR τ_{dust} are highly correlated,
385 however the correlation coefficient does not provide information on the τ_{dust} magnitude.
386 Therefore, the time series of MODIS and MISR τ_{dust} as the function of their coincident
387 days at a representative grid point (29.5 W, 19.5 N), as well as averaged within the $5 \times$
388 5 and 10×10 grid boxes centered at this point, are shown (Figure 6c-e). Again it is
389 found that the MODIS and MISR τ_{dust} are highly correlated on the daily basis, and it also

390 reveals that τ_{dust} is systematically greater for MISR than MODIS.

391 MISR τ_{dust} is larger than MODIS τ_{dust} over most of the tropical Atlantic when
392 averaged for coincident days (about 0.03 larger averaged over the basin), except in some
393 regions north of the Equator (Figure 7b). This difference can be traced back to the
394 difference between MISR and MODIS total AOT (Figure 7a). Over almost the entire
395 tropical Atlantic, coincident MISR AOTs are systematically larger than the MODIS ones
396 (up to 0.08, and about 0.04 when averaged for the Atlantic basin). This is consistent with
397 previous studies in which MISR AOT is found to be generally larger than MODIS AOT
398 over water [e.g., *Abdou et al.*, 2005; *Kahn et al.*, 2010]. Further examination of
399 coincident MODIS and MISR τ and τ_{dust} binned against the MISR τ reveals that larger
400 MISR τ is found for the entire τ spectrum, whereas MISR τ_{dust} is larger (smaller) when
401 the aerosol concentration is relatively low (high) (Figure 7c). Note that lower aerosol
402 concentrations are overwhelmingly more frequent (black line in Figure 7c), thus the
403 averaged MISR τ_{dust} is larger than the averaged MODIS τ_{dust} . As seen above, the τ_{dust}
404 difference between MODIS and MISR can be traced back to the τ difference between
405 them. However, this is not the only reason, as, unlike MISR τ , MISR τ_{dust} is not
406 systematically larger than MODIS τ_{dust} . The uncertainties involved in the derivation of
407 both MODIS and MISR τ_{dust} inevitably lead to their differences too. However it should be
408 noted that although both methods have limitations, MISR discrimination of aerosol type
409 is much more robust than MODIS discrimination. The derivation of MISR τ_{dust} is based
410 on actual retrieved particle property information, whereas for MODIS, the accuracy of
411 the formula is subject to the use of constant FMF values and empirically calculated τ_{mar} .
412 Despite more physically robust separation of τ_{dust} from τ in MISR, it has much less

413 frequent sampling compared to MODIS, thus it is necessary to examine both datasets.
414 Results based on these two complementary satellite datasets will provide more solid
415 insight to the characteristics of the MJO-related dust and to some extent smoke
416 variabilities.

417 **4. MJO-related Atlantic Dust and Smoke AOT Anomalies**

418 **4.1. MJO Analysis Methodology**

419 For the MJO analysis and composite procedure, we use the multivariate empirical
420 orthogonal function (EOF) method introduced by *Wheeler and Hendon* [2004] and
421 adopted widely by the MJO community [e.g., *Waliser et al.*, 2009]. Briefly, the intra-
422 seasonal anomalies of daily AOT are obtained by removing the climatological-mean
423 seasonal cycle and filtering via a 30–90-day band pass filter. Then, a composite MJO
424 cycle (8 phases) is calculated by averaging the daily anomalies that occur within each
425 phase of the MJO cycle. The MJO phase for each day is determined by the Real-time
426 Multivariate MJO (RMM) index (a pair of PC time series called RMM1 and RMM2).
427 Only days having strong MJO activity ($\text{RMM1}^2 + \text{RMM2}^2 \geq 1$) are considered.

428 Figure 8 shows the number of strong MJO days, or events
429 ($\text{RMM1}^2 + \text{RMM2}^2 \geq 1$), used for the 8-phase MJO cycle composite for both MODIS and
430 MISR during 2002–2009 boreal winters. The number of total strong MJO events for each
431 MJO phase during this period is also shown at the right upper corner of each panel. For
432 both instruments, the number of strong MJO events used for the composite is much less
433 than the actual number of total events because of satellite retrieval sampling issues and
434 the data rejection applied. Generally, the number of MODIS events ranges from about 10
435 to 45, about as three times more than MISR, due to the much wider MODIS swath.

436 4.2. MODIS and MISR

437 The 8-phase MJO composite maps of MODIS total τ , τ_{dust} and τ_{smoke} anomalies are
438 shown in Figure 9a, 10a and 11a, respectively. The MISR counterparts will be discussed
439 and compared with MODIS results later in Section 4.4. Comparing Figure 9a and 10a, it
440 is seen that MODIS τ and τ_{dust} have very similar temporal evolution and spatial patterns,
441 except that τ_{dust} anomalies are slightly smaller. The spatial correlation between τ and τ_{dust}
442 anomalies for 8 MJO phases is 0.89. For both τ and τ_{dust} , strong negative anomalies (as
443 large as about -0.04) are found over the entire equatorial Atlantic for MJO phases 1, 2, 3,
444 and 8. In contrast, strong positive anomalies (up to 0.04) are found over the equatorial
445 Atlantic for MJO phases 5–6. For MJO phase 4, strong positive anomalies occur to the
446 north of the equator, whereas negative τ anomalies are found to the south. The converse
447 is true for the MJO phase 7. The MJO composite maps of total τ greatly resemble those
448 shown by *Tian et al.* [2011], with spatial correlation 0.52 (significant at the 99.9% level),
449 and have very similar magnitudes. This resemblance suggests that the MJO-related total
450 AOT anomaly patterns are robust and not sensitive to the data sampling (fewer samples
451 are used in this study). The MJO-related τ_{smoke} anomalies in MODIS are very weak, rarely
452 exceeding 0.01(Figure 11a). The spatial correlation between τ_{smoke} and τ is only about
453 0.21. The MJO composite maps of τ_{mar} anomalies are not shown since τ_{mar} is linearly
454 dependent on the surface wind speed, thus the MJO-related τ_{mar} anomaly pattern in fact
455 reflect the wind anomalies associated with the MJO, which has been examined in *Tian et*
456 *al.* [2011]. Furthermore, it is found that the magnitudes of the τ_{mar} anomalies are
457 negligible: the strongest negative/positive anomalies are about -0.004/0.004 (figure not
458 shown).

459 The 8-phase MJO composite maps of MISR total τ , τ_{dust} and τ_{smoke} anomalies are
460 shown in Figures 9b, 10b and 11b. Again, it is found that the τ_{dust} anomalies are
461 significantly larger than the τ_{smoke} anomalies. The τ_{dust} anomalies have very similar
462 patterns to those of total τ , with slightly smaller magnitude, whereas the τ_{smoke} anomalies
463 are very small and noisy. Further comparison between the MISR and MODIS τ and τ_{dust}
464 results (Figures 9 and 10) indicates that overall they exhibit very similar temporal
465 evolution as well as anomaly patterns despite the systematically larger MISR anomalies
466 compared to those of MODIS, which is likely the outcome of the systematically larger
467 MISR τ and τ_{dust} retrievals compared to MODIS over the tropical Atlantic Ocean, as
468 discussed earlier. The rejection of all $\tau < 0.15$ data in MISR could contribute to the above
469 difference, too; however, MISR τ anomalies with no data rejection don't show evident
470 differences (Figure not shown). Besides the magnitude difference, MISR τ and τ_{dust}
471 anomalies are also noisier than the MODIS anomalies due to lower sampling.
472 Nevertheless, the consistency between the MODIS and MISR results shown in Figures 9-
473 11 demonstrate that dust is the dominant aerosol component on the intra-seasonal time
474 scale, and the MJO-related dust anomalies are robust, as seen from two independent sets
475 of satellite observations with different strengths and limitations.

476 **4.3. Sensitivity of MODIS Results**

477 *Kaufman et al.*'s method to compute τ_{dust} and τ_{smoke} using MODIS τ and f based
478 on equations (3) and (4) is straightforward; however, large uncertainties in the computed
479 τ_{dust} and τ_{smoke} are expected for at least several reasons: the uncertainties in the MODIS τ
480 and f retrievals, the uncertainties in empirically computed τ_{mar} , and most critically, the
481 uncertainties resulted from assuming constant f_{mar} , f_{dust} , and f_{smoke} . In this subsection, we

482 will examine the sensitivity of our results to the FMF values. We first perturb the FMF
483 values used in this study (see Section 2.1.1) by increasing/decreasing one of them at a
484 time by its uncertainty range while keeping the other two unchanged. These sensitivity
485 test cases are denoted group 1. We then test using FMF values derived by three other
486 studies [*Jones and Christopher*, 2007; 2011; *Yu et al.*, 2009], which are denoted group 2.
487 These studies have attempted to calibrate Kaufman's technique, and re-derived the FMFs
488 using either updated MODIS datasets, or aerosol observations from other satellite
489 datasets, or using the GOCART (Goddard Chemistry Aerosol Radiation and Transport)
490 model to locate regions where a single aerosol component dominates. The FMF values
491 they obtained are generally consistent with what is used in this study: f_{mar} , f_{dust} , and f_{smoke}
492 are 0.25, 0.44, and 0.83, respectively in the study by *Jones and Christopher* [2007], and
493 0.31, 0.49, and 0.78 in their 2011 study, and 0.45, 0.37, and 0.90 in *Yu et al.* [2009].
494 FMFs do vary, and these studies find that they vary considerably depending on region
495 and season.

496 Figure 12a shows the composite 8-phase MJO cycles of τ_{dust} anomalies averaged
497 over the equatorial Atlantic (30 W–15 W, EQ–15 N) based on different sensitivity test
498 cases. The 30 W–15 W, EQ–15 N box is chosen to represent the region with strong
499 intra-seasonal aerosol variations. Overall, the colored lines (9 sensitivity tests) cluster
500 around the solid black curve (control case), and the MJO cycles of τ_{dust} anomalies based
501 on different sensitivity tests show a coherent evolution. This suggests that the MJO
502 associated τ_{dust} anomalies over the tropical Atlantic are quite robust despite the
503 uncertainties in using constant FMFs. Nevertheless, the lines spread in some phases. The
504 result based on *Yu et al.* [2009] is most different from other cases, probably because in

505 the other studies, the FMF values follow the sequence $f_{\text{mar}} < f_{\text{dust}} < f_{\text{smoke}}$ despite the
506 deviations, whereas the order of f_{mar} and f_{dust} is reversed in *Yu et al.*'s study [2009]. The
507 spread in group 2 is naturally larger than that in group 1. Furthermore, the τ_{dust} anomalies
508 are more sensitive to f_{dust} , as indicated by the larger deviations of the dashed cyan and
509 blue curves to the control case in phase 2, 3 and 6.

510 Similarly, Figure 12b shows the sensitivity test results for τ_{smoke} anomalies. The
511 spread of τ_{smoke} anomalies is quite large compared to their magnitude. Nevertheless, the
512 overall MJO cycle of τ_{smoke} anomalies is consistent among the different cases and their
513 magnitudes are much smaller than that of the τ_{dust} anomalies.

514 **5. Summary and Conclusions**

515 Previous studies [*Tian et al.*, 2008; 2011] found significant intra-seasonal
516 variability related to the MJO in the total column AOT over the tropical Atlantic region.
517 Aerosol over the tropical Atlantic is primarily a mixture of mineral dust, biomass burning
518 smoke, and marine aerosol. Given the different roles in the radiative forcing and cloud
519 formation processes played by these three aerosol types, as well as the potential
520 predictability of the MJO extending to 2–4 weeks, it is of great interest to further examine
521 the MJO-related variability for individual aerosol types, especially dust and smoke
522 aerosols.

523 Daily MODIS/Aqua total AOT and FMF measurements are used to derive daily
524 τ_{dust} and τ_{smoke} following *Kaufman et al.*'s method [2005a, b]. This method contains
525 considerable uncertainties, due to the assumption of constant f_{dust} , f_{smoke} , f_{mar} as well as
526 the empirical calculation of τ_{mar} . Strict data rejection has been applied in order to use
527 Kaufman's formula in a safer way. With MISR's sensitivity to aerosol particle shape,

528 dust and smoke aerosols can be distinguished using the MISR aerosol non-spherical
529 fraction. Because MISR sensitivity to shape (and MODIS sensitivity to FMF) diminishes
530 when aerosol concentration is low, only $\tau \geq 0.15$ data are used to compute τ_{dust} and τ_{smoke}
531 for MISR. The examination of both datasets is necessary, and results from MODIS and
532 MISR are complementary: MODIS provides better spatial coverage, and MISR dust is
533 derived from actual aerosol property retrieval rather than assumed aerosol-type-specific
534 FMF.

535 Both MODIS and MISR show a very similar dust and smoke winter climatology.
536 τ_{dust} is found to be the dominant aerosol component over the tropical Atlantic. It is largest
537 over the eastern equatorial Atlantic (about 0.35) and gradually decreases toward the west.
538 τ_{smoke} is significantly smaller than τ_{dust} . Over the equatorial Atlantic with strong total
539 aerosol loading, the contribution of τ_{dust} to total τ ranges from 40% to 70%, in contrast to
540 about 25% is attributed to τ_{smoke} and less than 20% from τ_{mar} . The daily MODIS and
541 MISR τ_{dust} are overall highly correlated, with the correlation coefficients typically about
542 0.7 over the North Atlantic, but much smaller or even negative over the South Atlantic.
543 MISR τ_{dust} is found to be systematically greater than the coincident MODIS τ_{dust} , and this
544 difference can be traced to the AOT difference between them. The consistency of the
545 MODIS and MISR dust and smoke aerosol climatology and daily variations give us
546 confidence to use these two data sets to investigate the relative contribution of dust and
547 smoke aerosols to the total AOT variation associated with the MJO. However, the
548 identification of smoke is much less certain than that of dust, because discrimination
549 among fine-mode sea salt, sulfate, and smoke particles depends on assumptions for both
550 MODIS and MISR, whereas the MISR dust discrimination is based on retrieved particle

551 shape.

552 For MODIS, the MJO composite maps of τ_{dust} anomalies are very similar to those
553 of τ anomalies, and are of comparable magnitude. Furthermore, the variance of τ_{dust}
554 anomalies on the full intra-seasonal time scale is found to be comparable or even bigger
555 than that of the τ anomalies. In contrast, the MJO-related τ_{smoke} anomalies are rather
556 small, barely exceeding 0.01, and the τ_{mar} anomalies are negligible. The sensitivity study
557 further shows that the MJO-related τ_{dust} and τ_{smoke} anomalies are quite robust, even when
558 we perturb the FMFs by their uncertainty ranges or use different sets of FMFs from
559 several independent studies.

560 Similarly, MISR also shows that the MJO composite maps of τ_{dust} anomalies are
561 very similar to those of τ anomalies, while the MJO-related τ_{smoke} anomalies are rather
562 small. The composite MJO cycle of τ_{dust} anomalies from MISR over the tropical Atlantic
563 Ocean is consistent with the MODIS results although the anomalies are much noisier due
564 to its less frequent sampling. The magnitude of MISR anomalies is again found to be
565 systematically larger than that of MODIS. The MJO-related τ_{smoke} anomalies in MISR are
566 overall slightly larger than for MODIS, but still much smaller compared to MISR τ_{dust}
567 anomalies.

568 The consistency between the MODIS and MISR τ_{dust} and τ_{smoke} anomalies in
569 terms of the evolution of the MJO cycle as well as the spatial pattern of anomalies
570 suggests that dust aerosol is the dominant component on the intra-seasonal time scale
571 over the tropical Atlantic Ocean.

572 The observational results obtained from two complementary satellite datasets can
573 be used to evaluate chemical transport models and help in model development.

574 Furthermore, given the potential predictability of the MJO, this study implies the MJO
575 modulation of Atlantic dust aerosol concentration may be predictable with lead times of
576 2–4 weeks, which in turn may lend guidance to many other dust-related phenomena, such
577 as the air quality, dust storm activity, and ocean nutrient deposition over the Atlantic
578 Ocean.

579

580 **Acknowledgments**

581 This research was performed at Jet Propulsion Laboratory (JPL), California
582 Institute of Technology (Caltech), under a contract with National Aeronautics and Space
583 Administration (NASA). It was supported in part by the National Science Foundation
584 (NSF) grant ATM-0840755 at University of California, Los Angeles. The work of R.
585 Kahn is supported in part by NASA's Climate and Radiation Research and Analysis
586 Program, under H. Maring, NASA's Atmospheric Composition Program under R.
587 Eckman, and the EOS-MISR project. The MODIS/Aqua data used in this study have
588 been obtained from the NASA LAADS server and ERA-Interim data used in this study
589 have been obtained from the ECMWF Data Server. © 2011. All rights reserved.

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742 **Figure Captions**

743 **Figure 1:** Calculated a) τ_{dust} and b) τ_{smoke} as a function of total τ and f according to
744 equations (3) and (4). Here, $f_{\text{mar}}=0.3$, $f_{\text{dust}}=0.5$, $f_{\text{smoke}}=0.9$, and $\tau_{\text{mar}}=0.06$, as indicated by
745 four white straight lines. The white dashed lines indicate $\tau_{\text{dust}}=-0.03$ and $\tau_{\text{smoke}}=-0.03$.
746 Contour interval: 0.1.

747 **Figure 2:** a) The count of paired MODIS τ and f measurements with respect to τ and f
748 over the tropical Atlantic (20°S–30°N, 60°W–20°E) for 2002–2009 boreal winters.
749 Superimposed solid lines are the lines of $\tau_{\text{dust}}=-0.03$ and $\tau_{\text{smoke}}=-0.03$ in Figure 1. b) The
750 percentage of days with MODIS τ and f measurements during 2002–2009 boreal winters
751 (1269 days in total). c) The percentage of τ and f measurements used to calculate τ_{dust} and
752 τ_{smoke} relative to the total number of measurements.

753 **Figure 3:** a) The percentage of days with MISR τ measurements during 2002–2009
754 boreal winters (1269 days in total). b) The percentage of MISR $\tau \geq 0.15$ relative to total
755 number of τ measurements.

756 **Figure 4:** MODIS climatological mean (2002–2009) boreal winter a) total τ , b) τ_{dust} , d)
757 τ_{smoke} , and f) τ_{mar} over the tropical Atlantic Ocean, as well as their percentages to τ (c-
758 τ_{dust} , e- τ_{smoke} , and g- τ_{mar}). 9-point spatial smoothing is applied.

759 **Figure 5:** Same as Figure 3a-3e but for MISR, and only cases with MISR $\tau \geq 0.15$ used.

760 **Figure 6:** a) Correlation coefficients between coincident MODIS τ_{dust} and MISR τ_{dust}
761 during 2002-2009 over the tropical Atlantic and b) the number of coincident days. Time
762 series of MODIS (blue) and MISR (red) τ_{dust} as the function of the coincident day c) at
763 the point of 29.5 W, 19.5 N, averaged on the d) 5×5 and e) 10×10 boxes centered
764 at 29.5 W, 19.5 N.

765 **Figure 7:** a) MISR and MODIS τ difference averaged over coincident days during 2002-
766 2009. b) Same as a), but for τ_{dust} . c) Number of coincident MODIS and MISR
767 observations (black line), averaged MODIS τ (thick blue line), MISR τ (thick red line),
768 MODIS τ_{dust} (thin blue line), and MISR τ_{dust} (thin red line) over 20°S–30°N, 60°W–20°E
769 during 2002–2009 as a function of binned MISR τ (binned by every 0.01).

770 **Figure 8:** The number of strong MJO events used for MJO composite for each phase of
771 the MJO cycle for a) MODIS and b) MISR during the 2002–2009 boreal winters. The
772 total number of strong MJO events during this period is indicated at the right upper
773 corner of each panel.

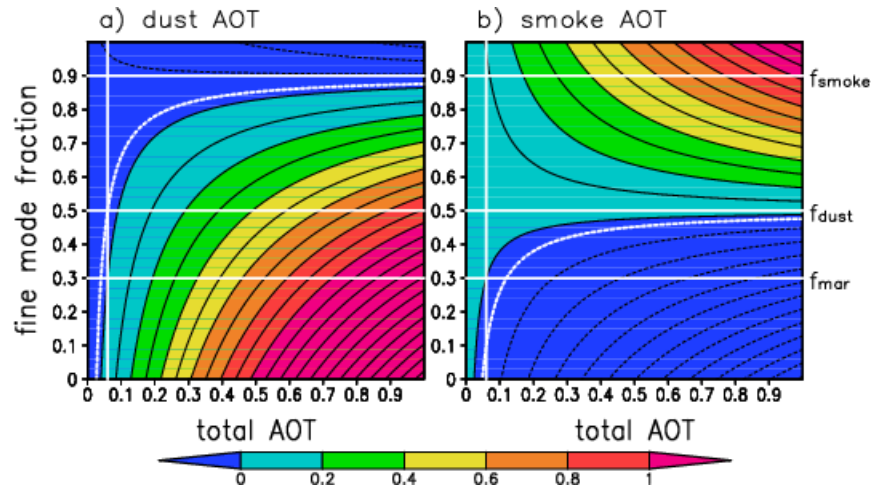
774 **Figure 9:** MJO composite maps of total τ anomalies (multiplied by 100) for a) MODIS
775 and b) MISR over the tropical Atlantic Ocean. 9-point spatial smoothing is applied.

776 **Figure 10:** Same as Figure 9, but for τ_{dust} .

777 **Figure 11:** Same as Figure 9, but for τ_{smoke} .

778 **Figure 12:** The composite MJO cycle of MODIS a) τ_{dust} and b) τ_{smoke} anomalies averaged
779 over 30 W–15 W, EQ–15 N for the control case (black) and 9 sensitivity test cases
780 (color) with f_{mar} , f_{dust} , and f_{smoke} of each case indicated in the legend.

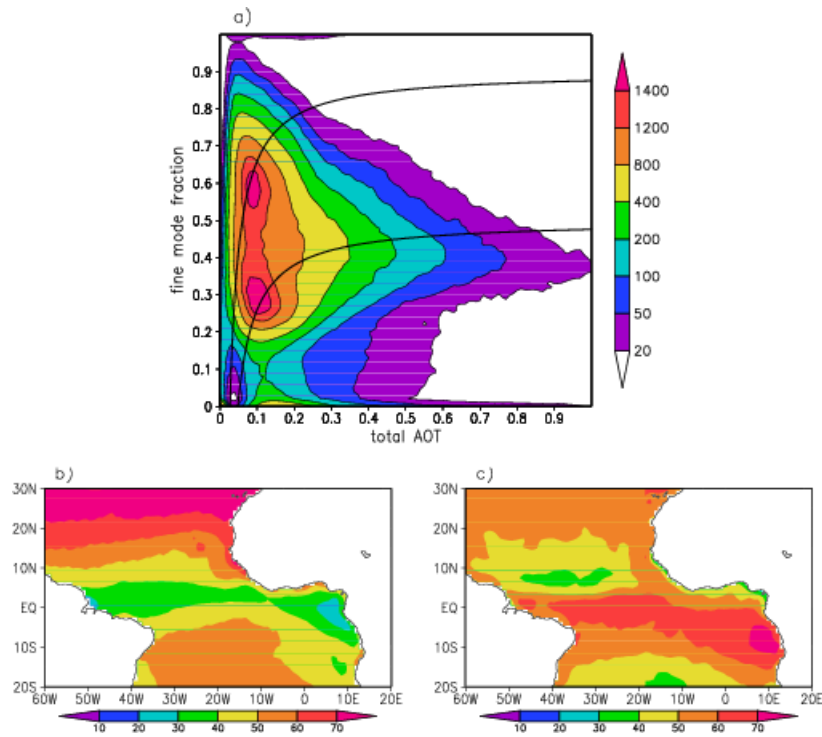
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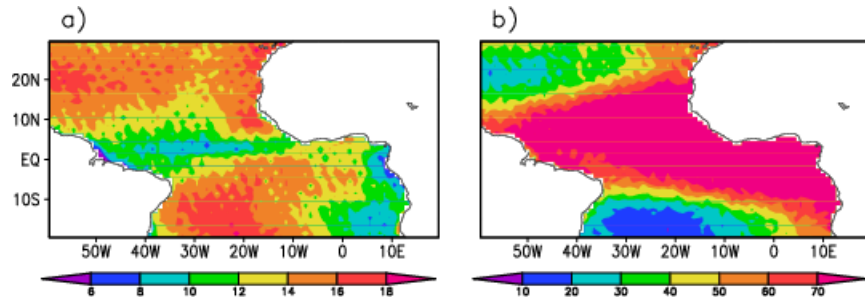


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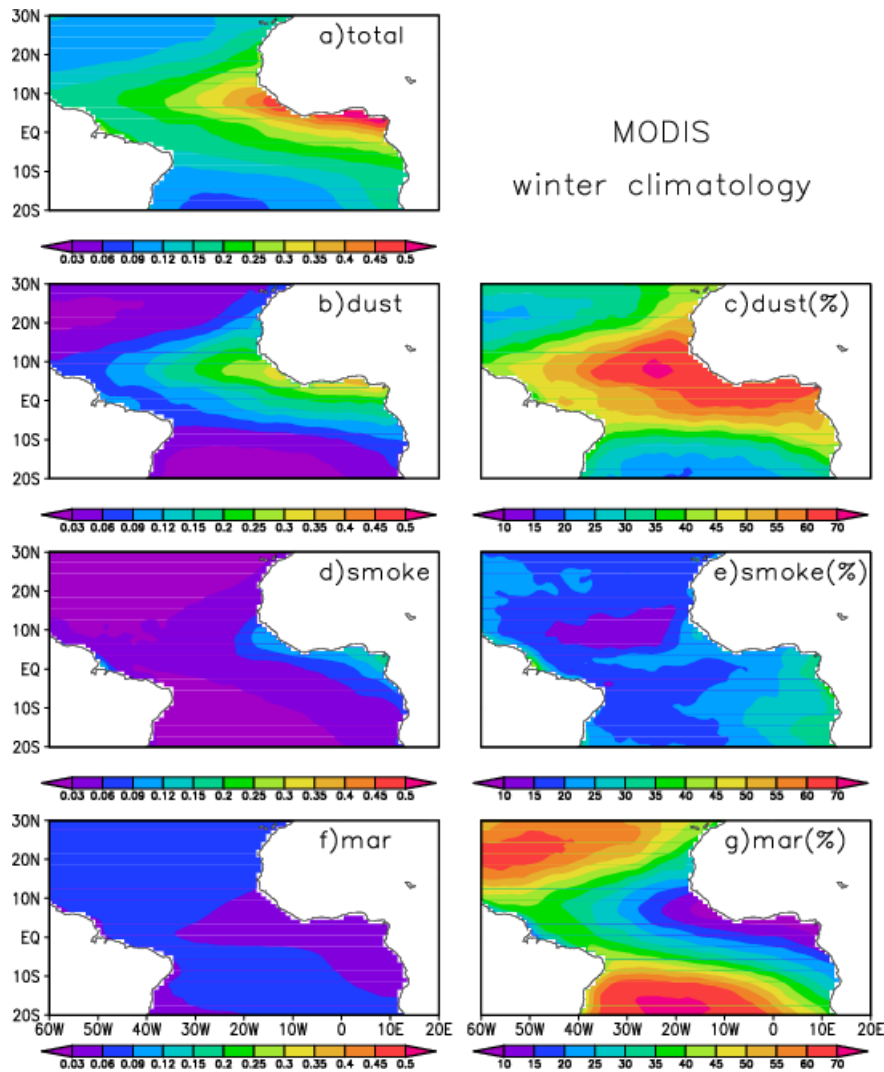
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798 **Figure 3:** a) The percentage of days with MISR τ measurements during 2002–2009
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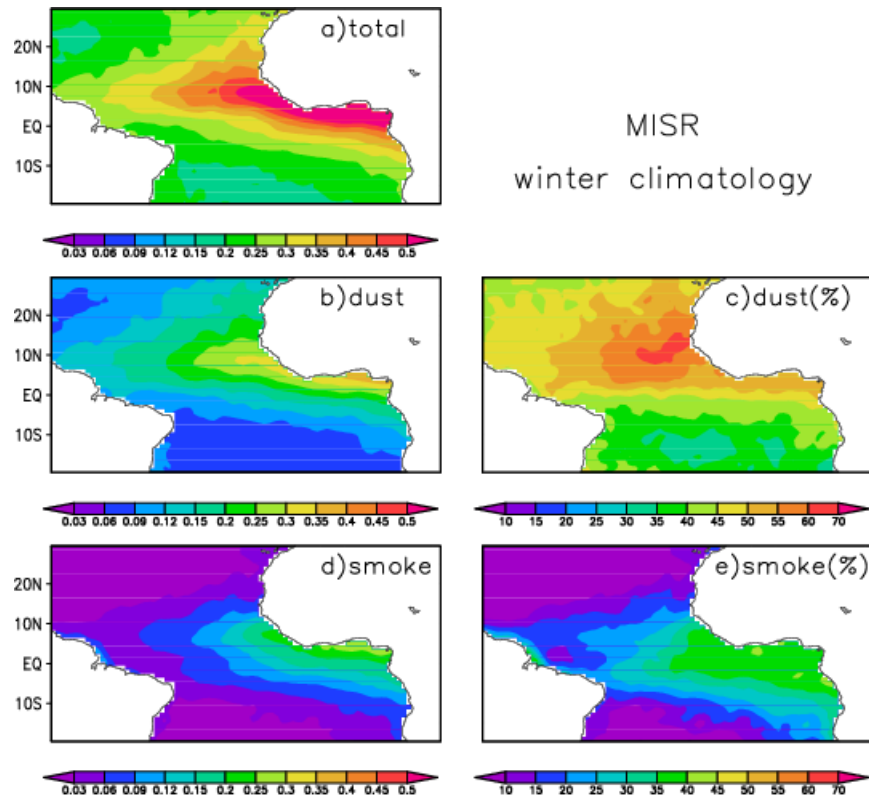
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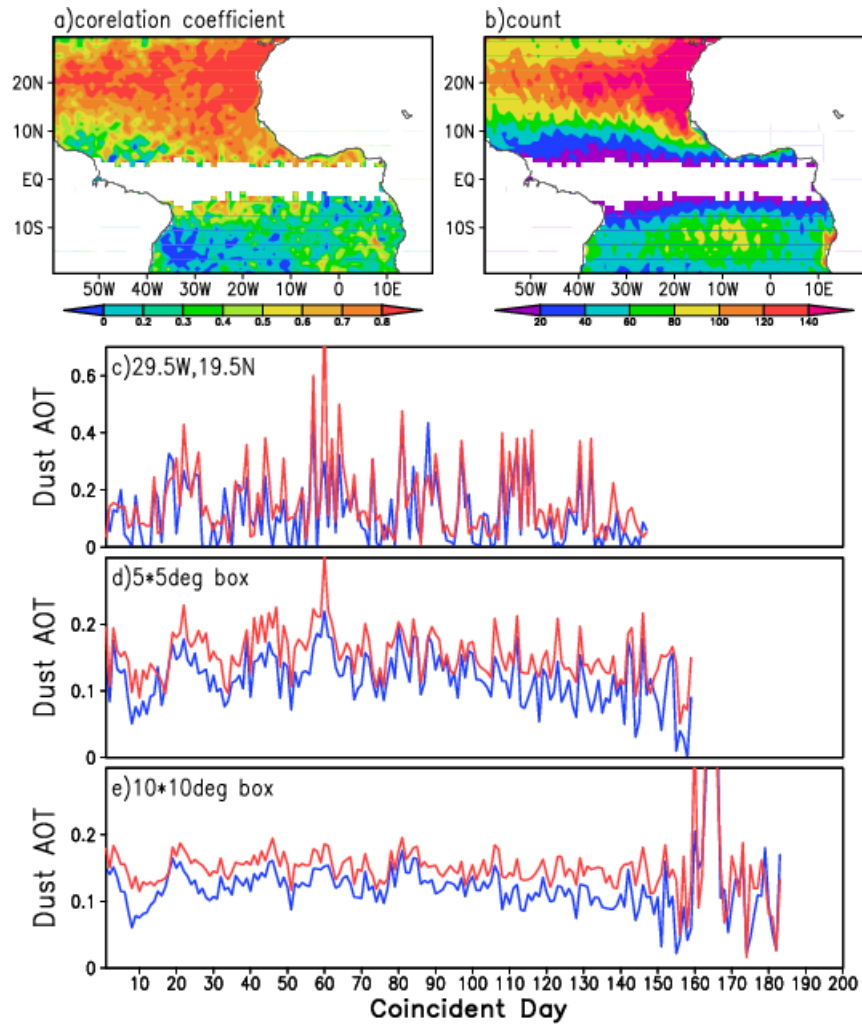
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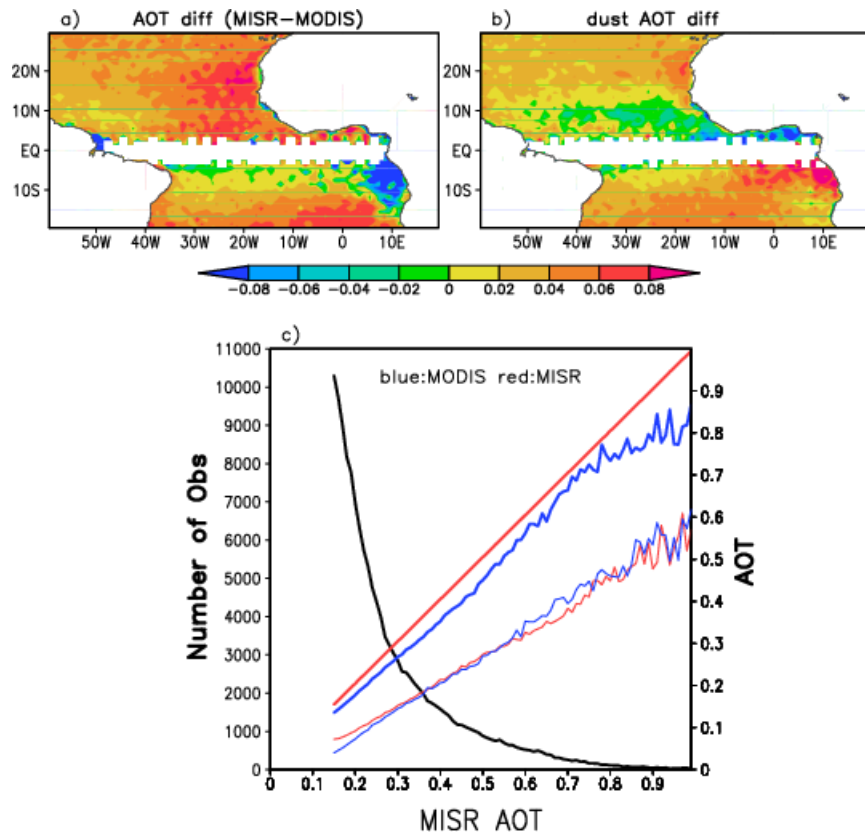
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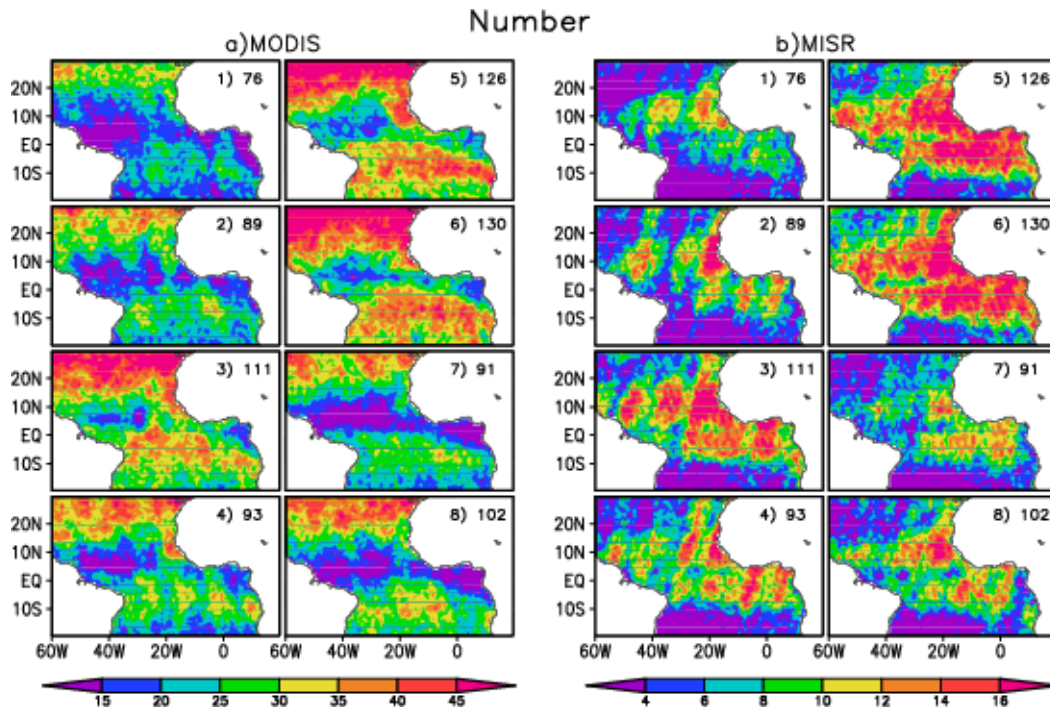
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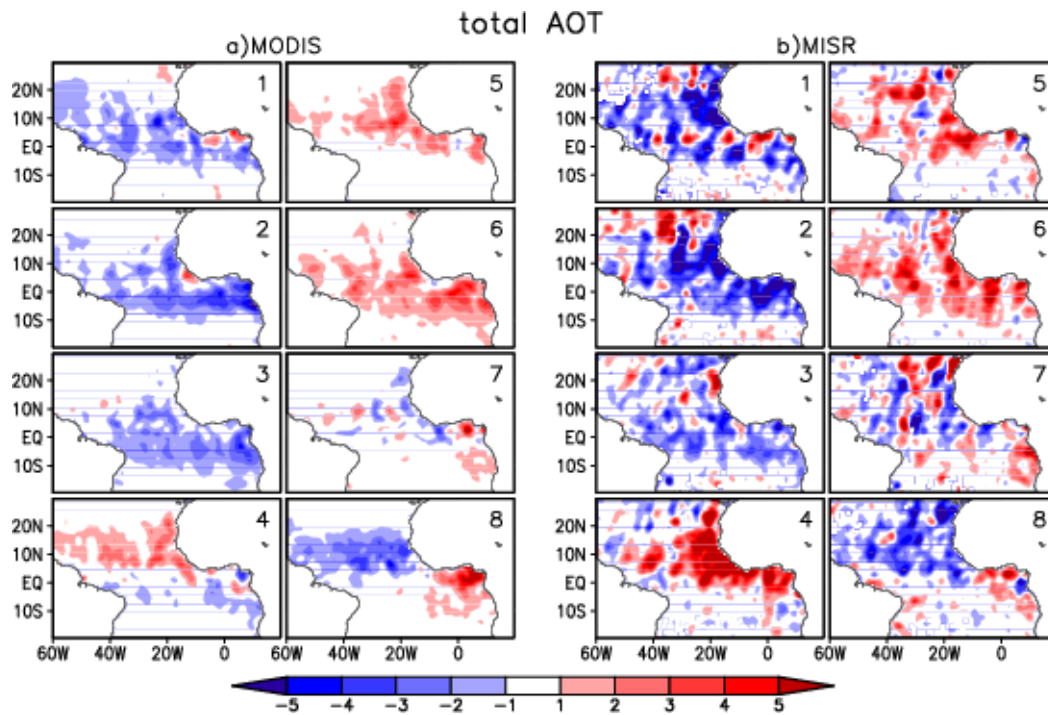
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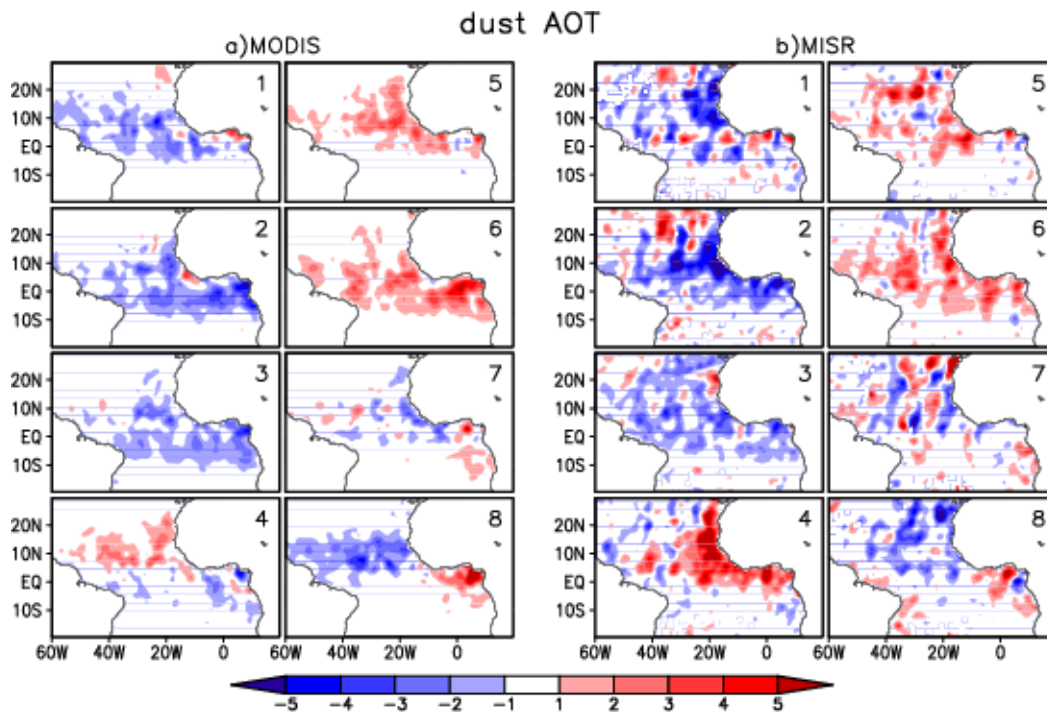
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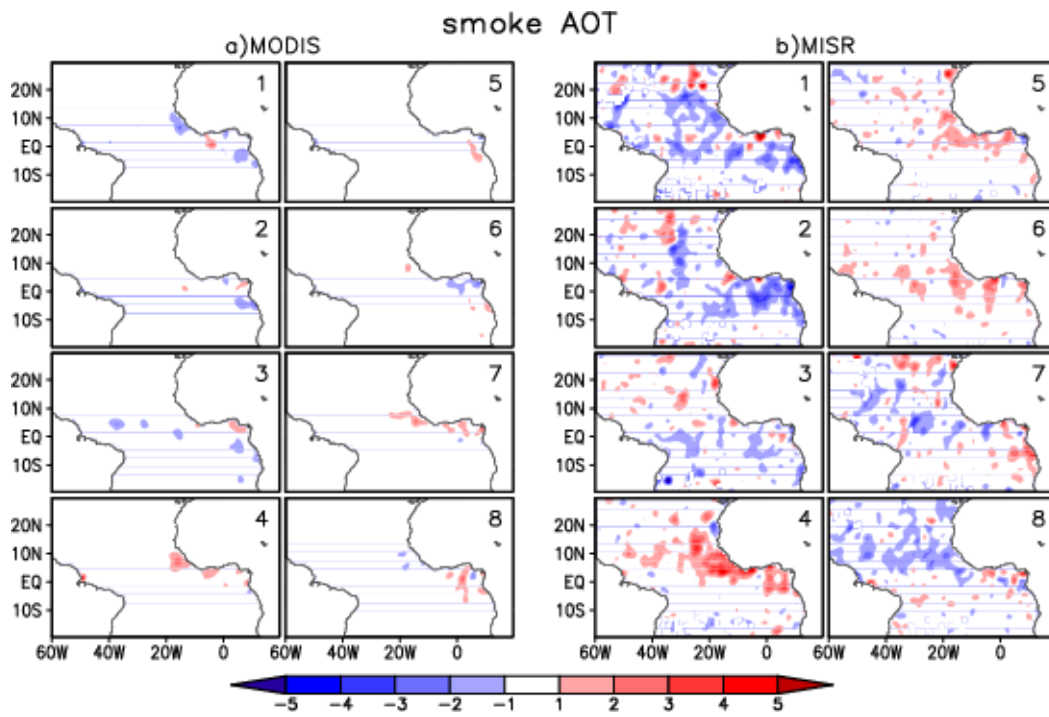


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831 **Figure 10:** Same as Figure 9, but for τ_{dust} .

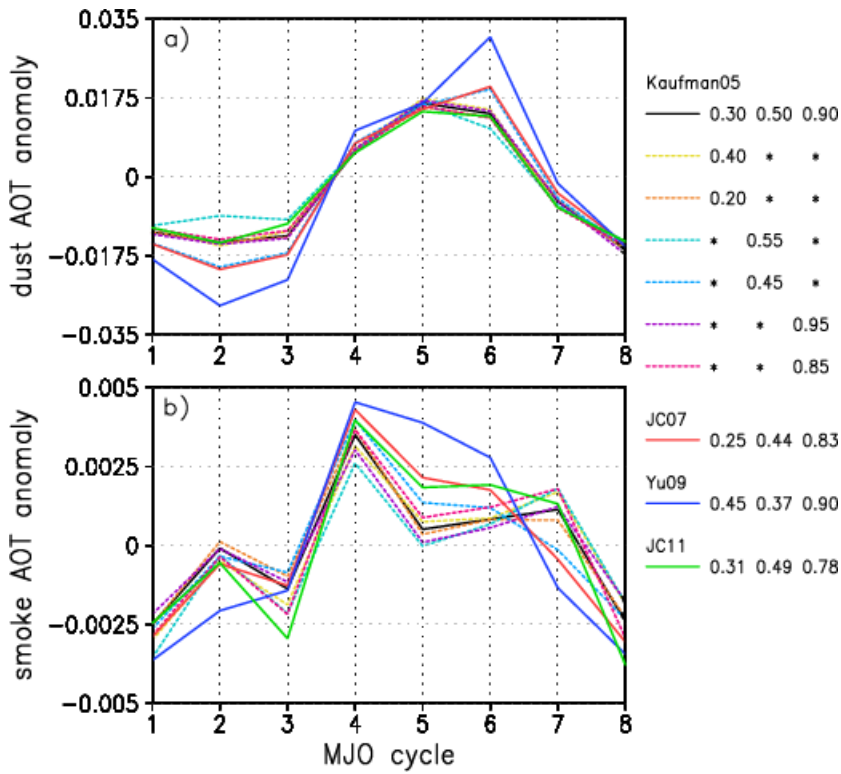
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