

1 **Effect of spectrally varying albedo of vegetation surfaces on shortwave**
2 **radiation fluxes and aerosol direct radiative forcing**

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Abstract

This study develops an algorithm for the representing large spectral variations of vegetation albedo based on Moderate Resolution Imaging Spectrometer (MODIS) observations at 7 discrete channels, referred to as MODIS Enhanced Vegetation Albedo (MEVA) algorithm. The MEVA algorithm empirically fills spectral gaps around the vegetation red edge near 0.7 μm and represents vegetation absorption features at 1.48 and 1.92 μm , which can't be adequately captured by the MODIS 7 channels. We then assess the effects of different characterizations of vegetation albedo (including MEVA and four traditional approaches to applying the MODIS observed discrete reflectance) on calculations of solar fluxes and aerosol direct radiative forcing (DRF) at the top of atmosphere (TOA). By comparing DRF results obtained through MEVA method to the results obtained through the four traditional approaches, we show that filling the reflectance gap of the MODIS measurements around 0.7 μm based on the general spectral behavior of healthy green vegetation leads to significant improvement in aerosol DRF at the top of atmosphere (TOA) (up to 3.02 Wm^{-2} being about 90% of the aerosol DRF calculated with surface reflectance of high spectral resolution) ; the corrections to the other two spectral gaps in the vegetation spectrum missed by the MODIS reflectances also contribute to improving TOA DRF calculations but to a much lower extent (less than 0.27 Wm^{-2} being about 57% of the DRF calculated with surface reflectance of high spectral resolution). Compared to traditional approaches, MEVA improves the accuracy of the outgoing solar flux at the top of the atmosphere by over 60 Wm^{-2} and aerosol DRF by over 10 Wm^{-2} in the tested cases. Specifically, for Amazon vegetation types, MEVA can improve the accuracy of daily averaged aerosol radiative forcing at equator at equinox by 3.7 Wm^{-2} . These improvements indicate that MEVA can contribute to

49 vegetation covered regional climate studies, and help to improve understanding of climate
50 processes and climate change.

51

52 **1 Introduction**

53 Vegetation covered land surface and the climate are linked together through complex ecological,
54 hydrological, and biogeochemical processes (Dickinson, 1983; Dirmeyer and Shukla, 1994;
55 Dickinson, 1995; Lyapustin, 1999; Betts, 2000; Lucht et al., 2002). Among these processes, the
56 surface directly reflects the solar radiation and affects the Earth's energy balance, and hence the
57 climate (Cess, 1978; Lofgren, 1995). The knowledge of the surface albedo properties affects
58 earth-atmosphere system related calculations and retrievals such as the direct aerosol forcing
59 calculation (Yu et al., 2006; McComiskey et al., 2008) and cloud properties retrieval (Popp et al.,
60 2011). Specifically, spectral surface albedo is affected by leaf structure, water content, pigment,
61 chlorophyl, etc. (Collins, 1978; Kim et al., 1994; Asner et al., 2000; Ceccato et al., 2001). The
62 black sky albedo calculated from the Moderate Resolution Imaging Spectroradiometer (MODIS)
63 data shown in Fig. 1 indicates that vegetation albedo has large spatial and spectral variations.
64 Adequate representation of these variations is important for estimating radiative flux and aerosol
65 radiative forcing.

66 Much work has been done to capture vegetation surface reflectance and albedo. Directly, surface
67 albedo can be obtained from field measurements (Gilgen et al., 1995; Sellers et al., 1992; Hall
68 and Sellers, 1995). Leaf samples have also been collected and their reflectance has been
69 determined by spectrophotometers in the laboratory (Hosgood et al., 1994; Clark et al., 2007).
70 Additionally, remote sensing techniques have been widely used to determine surface albedo, for

71 instance the Airborne Visible/Infrared Imaging Spectrometer (AVIRIS) (Staenz et al., 1996),
72 and many satellite operations including, but not limited to: Global Ozone Monitoring Experiment
73 (GOME) (Kolemeijer et al., 2003), MEdium Resolution Imaging Spectrometer (MERIS)
74 (Muller, 2006), Polarization and Directionality of the Earth's Reflectances (POLDER) (Leroy et
75 al., 1997), Multiangle Imaging Spectroradiometer (MISR) (Diner, 2008), Advanced Very High
76 Resolution Radiometer (AVHRR) (Saunders, 1990), Visible Infrared Imager Radiometer Suite
77 (VIIRS) (Miller, 2002), and Moderate Resolution Imaging Spectroradiometer (MODIS) (Liang
78 et al., 1999; Lucht et al., 2000; Schaaf et al., 2002; Moody et al., 2005).

79 Specifically, the MODIS sensor on board the NASA polar satellites TERRA (1999 – present)
80 and AQUA (2000 – present) measures the reflected solar radiation at the top of the atmosphere
81 (TOA) which can be used to retrieve surface properties (Vermote et al., 1997). The MODIS land
82 science team has used the MODIS measurements to develop a series of surface albedo data
83 products, including MCD43C among others. In detail, MCD43C provides three spectrally
84 dependent parameters f_{iso} , f_{vol} , f_{geo} for calculating black sky albedo and white sky albedo when
85 combined with a BRDF model. These parameters are available at MODIS bands 1 to 7
86 (nominally centered at 0.47, 0.55, 0.67, 0.86, 1.24, 1.63, and 2.11 μm), and in the spectral ranges
87 of visible (0.3 to 0.7 μm), near infrared (0.7 to 5 μm), and total broadband (0.3 to 5 μm).

88 Satellite remote sensing techniques have the advantages of having larger spatial and longer
89 temporal coverage than in situ measurements. However, most satellite sensors can only measure
90 reflectance at certain narrow bands and have the drawback of inadequately characterizing
91 spectral variations as the example shown in Fig.2. Clearly MODIS spectral measurements don't
92 well capture the rapid increase of reflectance from 0.67 to 0.86 μm and the dips at 1.48 and 1.92

93 μm . For flux and aerosol forcing calculations, broadband albedo is generally used and narrow
94 band albedo is usually ignored (Myhre et al., 2005; Zhou et al., 2005; Patadi et al., 2009). The
95 limited spectral reflectance information and the simplified broad band albedos can be error prone
96 in radiative forcing calculations (Wang et al., 2011).

97 This work presents a new algorithm – the MODIS enhanced vegetation albedo (MEVA) – to
98 provide an integrated vegetation reflectance spectrum, with the advantage of global and temporal
99 coverage over the lifetime of MODIS. Given reflectance at MODIS bands 1-7, this project
100 demonstrates that the vegetation reflectance spectrum determined by the MEVA algorithm
101 improves the accuracy of the TOA flux and aerosol forcing calculations.

102 **2 Methodology**

103 **2.1 Traditional approaches**

104 Several methods have been traditionally used to integrate the surface albedo over the whole solar
105 spectrum based on the MODIS bands 1-7. These methods are illustrated in Fig. 3 based on the
106 reflectance spectrum of *miconia guianensis* adapted from Arai et al. (2010) and can be described
107 as: (a) the narrowband reflectance is converted to reflectance in total shortwave broadband (from
108 0.3 to 2.5 μm) (Liang et al., 1999); (b) narrowband reflectance at MODIS bands 1-7 is converted
109 to reflectance in two broad bands: broadband “visible” (from 0.3 to 0.7 μm) and “near infrared”
110 (from 0.7 to 2.5 μm) according to Liang et al. (1999); (c) the wavelength between two adjacent
111 MODIS channels are averaged (which leads to 0.51, 0.61, 0.77, 1.10, 1.44, and 1.87 μm) and
112 seven reflectance values from MODIS centered in the native wavelengths are assigned to the
113 following bands: from 0.3 to 0.51 μm , from 0.51 to 0.61 μm , and so on (denoted “average band
114 MODIS” in following discussions); (d) the reflectance at MODIS bands 1-7 is linearly

115 interpolated. The approach in methods (a) and (b) is performed through polynomial regressions
116 to convert albedos at MODIS narrow bands to broadband albedos at visible, near infrared, and
117 total shortwave as described in Liang et al. (1999). Using the reflectance spectrum of vegetation
118 *miconia guianensis* adopted from Arai et al. (2010) as an example, Fig. 3 illustrates results from
119 the four approaches described above.

120 This research will show that all of these traditional techniques produce significant errors in
121 estimating TOA radiative fluxes and aerosol forcing. The new methodology proposed here based
122 on MODIS bands 1-7 (MEVA – MODIS enhanced vegetation albedo) will minimize these
123 errors. TOA solar fluxes and aerosol direct radiative forcing will be calculated for all these
124 methods (traditional and proposed) and will be compared with the results calculated from the
125 high resolution spectral libraries.

126 As shown by the solid blue line in Fig. 3(d), it is possible to linearly connect reflectance at
127 MODIS bands 1-7 in order to interpolate the reflectance data. However, in this method, there are
128 three distinct features missing from the actual spectrum, which can be seen in the shadowed
129 areas in Fig. 4 (b).

130 The first missing feature is associated with the vegetation red edge around 0.7 μm , which is the
131 division between the low reflectance in the visible and high reflectance in the near infrared. The
132 red edge in the vegetation surface reflectance spectra have been used to study chlorophyll, water
133 content, pigment content properties, and more (Horler et al., 1983; Guyot et al., 1992; Gitelson et
134 al., 1996; Sims and Gamon, 2002; Stimson et al., 2005). As shown in Fig. 4 (a), solar radiation
135 arriving at the surface is relatively strong around 0.7 μm , which intensifies the errors in flux and
136 aerosol forcing calculations associated with the missing feature of the red edge. The other two

137 important missing features in the interpolated spectrum are around 1.44 and 1.92 μm due to
138 radiation absorption by vegetation water. These two missing features are expected to lead to
139 smaller errors, due to the weaker solar radiation arriving at the surface resulting from the strong
140 atmospheric water vapor absorption in these two spectral ranges as shown in Fig. 4 (a). These
141 errors are discussed in more detail in the following section through flux and aerosol forcing
142 calculation in different spectral ranges.

143 **2.2 MODIS enhanced vegetation albedo (MEVA)**

144 We now describe a new empirical method which will show how the seven MODIS narrowband
145 albedos can be extended in a continuous reflectance spectrum to minimize errors in the
146 calculation of fluxes at the TOA and lead to more accurate aerosol radiative forcing and flux
147 calculations.

148 The MODIS enhanced vegetation albedo (MEVA) algorithm is proposed here to minimize the
149 errors in flux and aerosol forcing calculations associated with these missing features. In addition
150 to MODIS 7 channels, MEVA includes 7 auxiliary channels (0.69, 0.72, a variable channel at the
151 top of the red edge, 1.44, 1.84, 1.92, and 3 μm). Four of these auxiliary channels are shown in
152 Fig. 5: the reflectance at 0.69 μm is obtained by linearly extrapolating the reflectance at 0.55 and
153 0.67 μm ; the reflectance at 0.72 μm is the average between 0.69 μm and 0.86 μm ; the reflectance
154 at 1.44 μm is 40% of the reflectance at 1.24 μm ; the reflectance at 1.92 μm is 20% of the
155 reflectance value at 1.63 μm . The remaining three auxiliary channels are a variable channel at the
156 top of the red edge, and at 1.84 and 3 μm as shown in Fig. 6. The variable channel at the top of
157 the red edge is defined as the crossing point between the linearly extrapolated line connecting
158 0.69 to 0.72 μm and the linearly extrapolated line connecting 1.24 and 0.86 μm ; the reflectance

159 at 1.84 μm is determined by linearly interpolating the reflectance at 1.63 and 2.11 μm ; the
160 reflectance at 3 μm is set to zero. Finally, the reflectance between 0.3 and 0.4 μm were set
161 constant to the reflectance at 0.47 μm . The auxiliary channels and the values of ratios were
162 determined here by the general behavior of vegetation spectra including the vegetation red edge
163 associated with ChlorophyllII absorption at 0.7 μm and vegetation water absorption features at
164 about 1.5 and 1.9 μm (Hoffer [1987]). The final result is a reflectance spectrum based on the
165 MODIS bands 1-7 that better resembles the most important features of a typical vegetation
166 spectral reflectance. With *miconia guianensis* (named “vegetation 5” in the discussion) as an
167 example, the MEVA spectrum is displayed as the solid blue line in Fig. 6.

168 **3 Evaluation of the methodology**

169 In order to evaluate the relative merits of the MEVA methodology versus traditional approaches
170 to interpolate the MODIS bands 1-7, several high-resolution vegetation spectra from the
171 literature were used as input to the Santa Barbara DISORT Atmospheric Radiative Transfer
172 (SBDART) program (Ricchiuzzi et al., 1998) to calculate examples of the outgoing flux at TOA
173 and the direct aerosol forcing. Here the direct radiative forcing is defined as the difference of
174 total outgoing flux at TOA under clear sky with and without aerosols. A positive direct aerosol
175 forcing value indicates that aerosols warm the earth-atmosphere system, and a negative value
176 shows that aerosols cool the earth-atmosphere system.

177 The SBDART is a radiative transfer model based on the discrete ordinate method which includes
178 aerosols, gases, and surface properties (Ricchiuzzi et al., 1998) and can run with different
179 atmospheric input settings and customized spectral surface albedo. For this study, the main input
180 parameters are the spectral surface albedo in a spectral range of 0.3 μm to 2.5 μm with a 0.01 μm

181 resolution; the spectral aerosol single-scattering albedo (SSA), aerosol optical depth (AOD), and
182 phase function in a spectral range of 0.3 μm to 2.5 μm with a resolution of 0.1 μm to 0.2 μm ; and
183 the standard tropical atmospheric profile. The phase function was represented by 128 terms of
184 Legendre moments calculated with Mie theory based on the Amazonian aerosol model presented
185 by Dubovik et al. 2002. The outputs are the flux at TOA in $\text{Wm}^{-2} \mu\text{m}^{-1}$ from 0.3 to 2.5 μm with a
186 default resolution of 0.005 μm .

187 During the simulation, the surface is assumed to be Lambertian and albedo is equal to
188 reflectance. This approximation makes significant errors for directional radiance calculations but
189 not so for the total TOA flux and aerosol forcing calculations. In addition, the MODIS surface
190 albedo product already considers the surface BRDF effect. Thus we consider this approximation
191 to be acceptable for this study. We also assume that the surface albedo doesn't depend on solar
192 zenith angle. This assumption doesn't introduce significant errors to the calculation of daily
193 averaged aerosol radiative forcing (Yu et al., 2004).

194 One possible scenario for biomass burning aerosols over the Amazon region is studied here using
195 the following input parameters: AOD (at 0.55 μm) = 0.32 and 0.64, SSA (at 0.55 μm) = 0.89,
196 and solar zenith angle (SZA) = 30 degrees. Cases with different AODs, SSAs, and SZAs were
197 also studied, and will be discussed in the next section. The vegetation reflectance spectra used in
198 this study (denoted "true" in following discussions) were taken from the JHU spectral library, the
199 USGS Digital Spectral Library (Clark et al., 2007), and from the spectral signatures of leaves
200 from Amazonian trees presented by Aria et al. (2010). The MODIS data has 500 m spatial
201 resolution and might contain mixtures of different land and vegetation types. This is a limitation
202 of this study which uses the laboratory measurement of the leaf spectral reflectance as the land

203 surface albedo in radiative transfer simulations. Since one main application of MEVA in this
204 study is the calculation of biomass burning aerosol forcing and TOA flux calculation over the
205 Amazon where green vegetation dominates over the whole year, this assumption is appropriate.

206 **3.1 Dry grass, green grass, conifer, and deciduous surfaces**

207 In this section, the studied vegetation types are dry grass, green grass, conifer, and deciduous;
208 their reflectance spectra are provided by the JHU spectral library. Given reflectance at MODIS
209 bands 1-7, the reflectance spectra are reconstructed through the methods discussed in the last
210 section. Figure 7 shows the reflectance spectrum provided by the spectral library (“true”) and
211 MEVA. Though the dry grass case shows large difference between MEVA and the “true”
212 spectrum, it was kept in all our calculations as an example of the “worst case” scenario and to
213 demonstrate that even in this situation the fluxes and forcing errors are reasonable under control.

214 For each vegetation type, the outgoing solar flux at TOA and aerosol direct radiative forcing
215 (from 0.3 to 2.5 μm) were calculated with the reflectance spectrum obtained from the high
216 resolution spectral libraries (“true”), traditional approaches (the linear MODIS, the averaged
217 band MODIS, the Liang visible and near infrared, the Liang shortwave), and the MEVA method.
218 The results are summarized in Table 1 and Table 2. As shown in Table 1, MEVA produces the
219 outgoing flux that is closest to the “true” in all cases, with the difference varying from 0.58 to
220 1.31 Wm^{-2} , while the maximum deviation associated with other methods reaches 23 Wm^{-2} . A
221 simplistic and naive expectation would be that the surface does not matter to aerosol forcing
222 because the difference between two radiative fluxes would cancel out the impact of the surface
223 reflectance. However this cancellation does not happen generally, because aerosol forcing
224 depends on the balance among aerosol absorption, aerosol scattering, and surface reflectance.

225 For example, some aerosols can have a cooling effect over low reflectance surfaces (e.g.
226 vegetation), but have a warming effect over high reflectance surfaces (e.g. snow). With the
227 exception of dry grass, Table 2 shows that MEVA yields the aerosol forcing that is closest to the
228 “true” case with regard to aerosol forcing magnitude (varying from 0.43 to 0.63 Wm^{-2}) and
229 percentage (below 10 %). The exception of dry grass is explained in the following sensitivity
230 discussion.

231 The sensitivity of aerosol DRF in different spectral ranges are investigated and shown in Table 3
232 for a deciduous vegetation surface. For the aerosol forcing in the spectral range of 0.55 to 1.24
233 μm , MEVA provides a difference of 0.35 Wm^{-2} from the “true”, as compared to differences
234 between -1.41 to - 2.67 Wm^{-2} from traditional approaches. This demonstrates that MEVA
235 surpasses traditional approaches in calculating aerosol forcing. It can also be observed from
236 Table 3 that the spectral range from 0.55 to 1.24 μm presents the largest difference between each
237 method and calculations with the “true” spectrum (except for the case of Liang shortwave).
238 These results indicate the importance of filling the spectral gaps for the missing feature of the
239 vegetation red edge around 0.7 μm , which is consistent with the discussion in Sect.2.2. In the
240 same fashion, gap filling through MEVA for the other missing features around 1.48 and 1.92 μm
241 lead to the closest aerosol forcing to “true”, with differences of -0.03 and 0.01 Wm^{-2} , compared
242 to the values from 0.01 to 0.08 Wm^{-2} estimated through traditional approaches.

243 According to Table 3, the aerosol forcing differences for the gap filling of the vegetation water
244 absorption missing features (shown in the spectral range of 1.24 to 1.63 and 1.84 to 2.1 μm) are
245 smaller than the aerosol forcing difference obtained for the gap filling of the red edge missing
246 feature (shown in the spectral range of 0.55 to 1.24 μm), i.e. 0.03 and 0.01 being smaller than

247 0.35; 0.01 and 0.08 being smaller than 1.76; etc. This result indicates that gap filling for the
248 vegetation water absorption missing features has a relatively small impact on aerosol forcing
249 calculation than the impact from the missing red edge. This conclusion is well explained by the
250 relatively weaker solar radiation and stronger atmospheric water absorption around 1.48 μm and
251 1.92 μm than those around 0.7 μm as shown in Fig. 4 (a). This suggests that the results are not
252 sensitive to the percentages we proposed in Figure 5. Very similar results were derived in the
253 analysis of green grass, conifer, aspens, and Amazon vegetation. The spectral analysis for dry
254 grass indicates that the aerosol forcing difference of 1.84 Wm^{-2} between MEVA and “true” (as in
255 Table 2) is predominantly caused by the difference in the spectral range of 0.3 to 0.55 μm , where
256 the difference is 1.86 Wm^{-2} (compared to 0.17, -0.14, -0.01, -0.01, and -0.01 in the other five
257 spectral ranges: 0.55 to 1.24, 1.24 to 1.63, 1.63 to 1.84, 1.84 to 2.1 and 2.1 to 2.5 μm). This
258 might be related with the distinct spectral feature of dry grass in the range of the 0.3 to 0.55 μm
259 (as shown in Fig. 7). In Fig. 7, the spectral reflectance for green grass, conifers, and deciduous is
260 characterized with a reflectance peak starting from about 0.5 μm and ends at about 0.7 μm which
261 is absent from dry grass. This distinct spectral behavior by dry grass might be caused by its low
262 chlorophyll and vegetation water content (Hoffer, 1978). In general, the results in Table 3 justify
263 the gap filling procedure by MEVA for the three missing features shown in Fig. 4.

264 **3.2 Aspen surfaces**

265 In this section, the above procedures are applied to the reflectance spectrum for aspen surfaces
266 provided by the USGS digital spectral library (Clark et al., 2007). “Aspen 1” (green leaf), “aspen
267 2” (green leaf), “aspen 3” (yellow-green leaf), and “aspen 4” (yellow leaf) were sampled in
268 Boulder, Colorado, USA, and their reflectances were measured by a laboratory spectrometer;

269 “aspen 5” was sampled in Yellowstone National Park, Wyoming, USA, and its reflectance
270 spectrum was retrieved from AVIRIS data; “aspen 6” was collected in Denver, Colorado, USA,
271 and its reflectance spectrum is the average of the three measured spectra. Figure 8 shows the
272 reflectance spectra from “true” and MEVA for these six different aspen surfaces.

273 The outgoing flux at TOA and aerosol forcing were calculated using these reflectance spectra as
274 surface albedo, and the results are shown in Tables 4 and 5. With the exception of aspen 4,
275 MEVA leads to the minimum difference to “true” for both flux and aerosol forcing compared to
276 traditional approaches. The aerosol forcing difference between MEVA and “true” is 0.61 Wm^{-2}
277 for aspen 4, which is greater than the difference of -0.36 Wm^{-2} from the average band MODIS
278 method. This might be related with the leaf color being “yellow”, which implies strong
279 reflectance in the range of 0.57 to $0.59 \mu\text{m}$ which can be seen in Fig. 8. Similar to the spectral
280 behavior of dry grass, the spectral behavior of aspen 4 might be caused by its low chlorophyll and
281 water moisture content (Hoffer, 1978). This indicates that MEVA works best for green
282 vegetation types, but still produces reasonable results for yellow leaves. Overall, MEVA
283 consistently improves the accuracy of the calculated outgoing flux at TOA and aerosol forcing.

284 **3.3 Amazon vegetation**

285 Results for Amazonian vegetation are specifically investigated in this section. The Amazon
286 forest plays a unique role in climate change (Shukla et al., 1990; Nobre et al., 1991). However,
287 Amazon vegetation reflectance data is scarce (Roberts et al., 1990; Arai et al., 2010). The
288 reflectance spectrum for six Amazon vegetation types from Arai et al. (2010) were shown in Fig.
289 9 overlaid with their MEVA spectra. The spectra from Arai et al. did not show the reflectance

290 value in the range of 1.35 to 1.45 μm and 1.85 to 1.95 μm , which were linearly connected in this
291 study to represent “true”.

292 In a more detailed analysis, the outgoing flux at TOA and aerosol forcing were calculated with
293 three different typical biomass burning aerosol models shown in Fig. 10: SSA (at 0.55 μm) =
294 0.95, 0.89, and 0.83, with AOD (at 0.55 μm) = 0.32. In Fig. 10, the SSA curves are simulated
295 from Mie code (Wiscombe, 1980) with 1.4589 as the real part of the refractive index and three
296 cases of spectrally constant imaginary refractive index equal to 0.0073, 0.0173, and 0.0273.
297 According to aerosol optical properties studies (e.g. Dubovik et al., 2002; Eck et al., 2003),
298 biomass burning aerosols have a relatively spectrally constant imaginary refractive index. This
299 simplification has also been applied to the biomass burning aerosol study by Procopio et al.
300 (2003). The size distribution was calculated through the Amazonian forest aerosol model by
301 Dubovik et al. (2002).

302 Aerosol forcing efficiency (defined as aerosol direct radiative forcing per unit AOD) results
303 calculated with SSA (at 0.55 μm) = 0.95 (noted as “model 1”), 0.89 (noted as “model 2”), and
304 0.83 (noted as “model 3”) for Amazon vegetation types are presented in Fig. 11. Other
305 parameters used in these calculations include AOD (at 0.55 μm) = 0.32 and SZA = 30 degrees.
306 The results in Fig. 11 shows that MEVA yields an aerosol forcing efficiency (in $\text{Wm}^{-2} \text{AOD}^{-1}$)
307 closest to that provided by “true” surface albedo. Using broadband shortwave albedo could
308 introduce an error of 10 to 50 $\text{Wm}^{-2} \text{AOD}^{-1}$. Similar to the earlier results for green grass, canopy,
309 deciduous, and aspens, the results for Amazon vegetation also indicate that the MEVA algorithm
310 leads to the best approximation to the “true” surface albedo spectra, regarding the accuracy of the
311 outgoing flux at TOA and the aerosol direct forcing.

312

313 Moreover, the differences of the aerosol forcing efficiency associated with different methods to
314 estimate reflectance spectrum are averaged over the six Amazonian vegetation types studied
315 here. The results in Fig. 12 indicate that the aerosol forcing efficiency calculated through MEVA
316 is the closest to that from “true” than the other traditional approaches discussed here. The same
317 conclusion is drawn from the studies with AODs equal to 0.64 and 1.28 (at 0.55 μm).

318 To assess the impacts of surface spectral albedo approximations on the daily average aerosol
319 forcing, we perform the aerosol forcing calculations at the equator and in equinox condition. The
320 surface albedo is assumed to be SZA independent. Figure 13 indicates that MEVA yields aerosol
321 forcing closest to that from “true” than traditional approaches for all vegetation types and under
322 all SZAs (especially when SZA is smaller than 60 degrees). The daily averaged aerosol forcing
323 was determined by the 24 hour average aerosol forcing. Table 6 shows that MEVA produces a
324 daily average aerosol forcing closest to that from “true” compared to traditional approaches for
325 all vegetation types. We also average the daily average aerosol forcing at equator at equinox over
326 the studied six vegetation types. The results, shown in the last column of Table 6, indicate that
327 MEVA is the best approximation to the “true” case using the high resolution surface reflectance
328 spectrum. The magnitude of the average aerosol forcing difference between MEVA and “true” is
329 about 0.05 Wm^{-2} , much smaller than -0.95 , -1.09 , -0.73 , and 3.80 Wm^{-2} calculated using
330 traditional approaches. The magnitude of the ratio of this difference to that from “true” is also
331 the minimum at -0.9% compared to 18.0% , 20.6% , 13.8% , and -71.8% estimated through
332 traditional approaches. A similar investigation was done with different aerosol models: SSA (at
333 $0.55 \mu\text{m}$) = 0.95 and 0.83 ; AOD (at $0.55 \mu\text{m}$) = 0.64 and 1.28 . Consistently, the results show that

334 MEVA yields average aerosol forcing closest to that from “true” compared with traditional
335 approaches.

336 **4 Conclusion**

337 In this research, a new approach called MEVA was developed to estimate the continuous
338 vegetation reflectance spectrum using the reflectance measurements acquired from MODIS
339 seven bands, namely 0.47, 0.55, 0.67, 0.86, 1.24, 1.63, and 2.11 μm . The approach enhances the
340 MODIS vegetation albedo product by characterizing large spectral variation features at 0.7, 1.44,
341 and 1.92 μm that are missing in the MODIS observations. Several sources of vegetation spectral
342 reflectance were used to evaluate the MEVA approach: the JHU spectral library (for dry grass,
343 green grass, conifer, and deciduous surfaces), the USGS digital spectral library (for aspen
344 surface), and measurements of six Amazon vegetation types. The gap filling for the missing red
345 edge feature at 0.7 μm is the most significant due to the strong solar radiation input in this
346 spectral range; the other two gap fillings for vegetation water absorption signatures are less
347 important, due to the weaker solar radiation and strong atmospheric water absorption.

348 Flux and aerosol forcing calculation results indicate that MEVA has significant advantages over
349 traditional approaches in accurately calculating radiative fluxes and aerosol radiative forcing. For
350 the studied cases with AOD (at 0.55 μm) = 0.32, MEVA improved the accuracy of the outgoing
351 flux at TOA by up to 60 Wm^{-2} (nearly 20% of the flux value derived from “true”), aerosol
352 forcing by up to 10 Wm^{-2} (about 70% of the forcing value derived from “true”), daily averaged
353 aerosol forcing at equator at equinox by up to 3.7 Wm^{-2} (about 70% of the forcing value derived
354 from “true”). A similar conclusion was drawn from parallel studies applying AOD (at 0.55 μm)
355 = 0.64 and 1.28. For aerosol forcing, MEVA led to errors less than 1 Wm^{-2} with the exception of

356 dry grass which produced an error of 1.84 Wm^{-2} . This greater error might be associated with
357 lower chlorophyll and water content of dry grass compared with the other discussed vegetation
358 types. The combination of MEVA results with our retrievals of SSA for biomass burning
359 aerosols (Zhu et al., 2011) will improve the estimate of radiative forcing and their impacts on
360 climate by providing more accurate flux and aerosol forcing calculations.

361 Our exercise in this study shows that MEVA can be employed to improve the accuracy of flux
362 and aerosol forcing calculations for vegetated surfaces. Particularly, with the publically available
363 global surface albedo data at MODIS seven channels, MEVA can be integrated into radiative
364 transfer calculations and contribute to regional climate studies over vegetated areas. In this study,
365 the MEVA algorithm validation used laboratory measurements of leaf reflectance as land surface
366 albedo in radiative transfer simulations. This work can be further improved with the analysis of
367 real remote sensing data where individual pixel might be composed of mixed different land and
368 vegetation types including yellow leaves.

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370

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583 Table 1. The calculated outgoing solar flux (in Wm^{-2} ; from 0.3 to 2.5 μm) at TOA over dry
 584 grass, green grass, conifer, and deciduous surfaces. Other parameters used include AOD (at 0.55
 585 μm) of 0.32, SSA (at 0.55 μm) of 0.89, and SZA of 30 degrees. Bold numbers represent the
 586 difference in flux associated with different approaches to estimate the surface reflectance
 587 spectrum.

Flux (in Wm^{-2})	Dry grass	Green grass	Conifer	Deciduous
True	362.17	237.18	234.99	251.21
MEVA	361.59	238.49	235.84	252.49
Linear MODIS	357.37	223.53	218.61	235.17
Averaged band MODIS	360.04	223.05	218.84	234.98
Liang visible and near infrared	352.63	221.03	218.42	235.96
Liang shortwave	363.12	213.79	212.33	229.96
Differences of flux (in Wm^{-2})				
MEVA - True	-0.58	1.31	0.85	1.28
Linear MODIS - True	-4.8	-13.65	-16.38	-16.04
Average band MODIS - True	-2.13	-14.13	-16.15	-16.23
Liang visible and near infrared - True	-9.54	-16.15	-16.57	-15.25
Liang shortwave - True	0.95	-23.39	-22.66	-21.25

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590 Table 2. The calculated instantaneous aerosol direct radiative forcing (DRF, in Wm^{-2} ; from 0.3 to
 591 2.5 μm) in clear-sky condition over dry grass, green grass, conifer, and deciduous surfaces. Other
 592 parameters used include AOD (at 0.55 μm) of 0.32, SSA (at 0.55 μm) of 0.89, and SZA of 30
 593 degrees. Bold numbers represent the difference in aerosol forcing associated with different
 594 approaches to estimate surface reflectance spectrum. The numbers in parentheses indicate the
 595 ratio of the absolute difference to the aerosol forcing calculated with “true.”

DRF (in Wm^{-2})	Dry grass	Green grass	Conifer	Deciduous
True	14.28	-8.25	-7.86	-6.28
MEVA	16.12	-7.62	-7.43	-5.7
Linear MODIS	13.43	-10.39	-10.54	-8.84
Averaged band MODIS	15.2	-10.38	-10.33	-8.7
Liang visible and near infrared	17.72	-9.88	-10.07	-8.06
Liang shortwave	27.03	-0.46	-0.74	2.66
Differences of DRF (in Wm^{-2})				
MEVA - True	1.84 (13%)	0.63 (8%)	0.43 (5%)	0.58 (9%)
Linear MODIS - True	-0.85 (6%)	-2.14 (26%)	-2.68 (34%)	-2.56 (41%)
Average band MODIS - True	0.92 (6%)	-2.13 (26%)	-2.47 (31%)	-2.42 (39%)
Liang visible and near infrared - True	3.44 (24%)	-1.63 (20%)	-2.21 (28%)	-1.78 (28%)
Liang shortwave - True	12.75 (89%)	7.79 (94%)	7.12 (91%)	8.94 (142%)

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598 Table 3. The calculated instantaneous aerosol DRF (in Wm^{-2}) in different spectral ranges over
 599 deciduous surface. Other parameters used include: AOD (at 0.55 μm) = 0.32; SSA (at 0.55 μm)
 600 = 0.89; and SZA = 30 degrees. Bold numbers represent the differences in aerosol forcing between
 601 applying specific approaches and “true” in specified spectral ranges.

DRF (in Wm^{-2})	0.3 - 0.55 μm	0.55 - 1.24 μm	1.24 - 1.63 μm	1.63 - 1.84 μm	1.84 - 2.1 μm	2.1 - 2.5 μm
True	-10.25	3.37	0.47	0.11	0	0.02
MEVA	-9.99	3.72	0.44	0.11	0.01	0.02
Linear MODIS	-11.13	1.61	0.55	0.1	0.01	0.01
Averaged band MODIS	-10.07	0.7	0.53	0.11	0.01	0.02
Liang visible and near infrared	-10.31	1.41	0.55	0.17	0.04	0.08
Liang shortwave	0.4	1.96	0.17	0.07	0.01	0.03
Difference of DRF (in Wm^{-2})						
MEVA - True	0.26	0.35	-0.03	0	0.01	0
Linear MODIS - True	-0.88	-1.76	0.08	-0.01	0.01	-0.01
Average band MODIS - True	0.18	-2.67	0.06	0	0.01	0
Liang visible and near infrared - True	-0.06	-1.96	0.08	0.06	0.04	0.06
Liang shortwave - True	10.65	-1.41	-0.3	-0.04	0.01	0.01

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604 Table 4. The calculated integration of the outgoing solar flux (in Wm^{-2} ; from 0.3 to 2.5 μm) at
 605 TOA over aspen surface. Other parameters used in the calculation include AOD (at 0.55 μm) of
 606 0.32; SSA (at 0.55 μm) of 0.89; and SZA of 30 degrees. Bold numbers represent the difference
 607 of the flux associated with applying different approaches to estimate the reflectance spectrum.

Flux (in Wm^{-2})	Aspen 1	Aspen 2	Aspen 3	Aspen 4	Aspen 5	Aspen 6
True	235.81	296.29	291.82	302.30	172.65	218.16
MEVA	234.74	292.37	284.38	300.04	174.18	219.67
Linear MODIS	219.98	277.69	273.17	296.36	164.42	204.73
Averaged band MODIS	220.86	282.53	273.93	298.44	163.87	204.91
Liang visible and near infrared	218.08	278.32	263.25	277.86	162.01	202.83
Liang shortwave	208.10	252.65	219.64	253.95	165.69	198.23
Differences of flux (in Wm^{-2})						
MEVA - True	-1.07	-3.92	-7.44	-2.26	1.53	1.51
Linear MODIS - True	-15.83	-18.60	-18.65	-5.94	-8.23	-13.43
Average band MODIS - True	-14.95	-13.76	-17.89	-3.86	-8.78	-13.25
Liang visible and near infrared - True	-17.73	-17.97	-28.57	-24.44	-10.64	-15.33
Liang shortwave - True	-27.71	-43.64	-72.18	-48.35	-6.96	-19.93

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610 Table 5. The calculated instantaneous aerosol DRF (in Wm^{-2} ; from 0.3 to 2.5 μm) over aspen
611 surfaces. Other parameters used in the calculation include AOD (at 0.55 μm) of 0.32; SSA (at
612 0.55 μm) of 0.89; and SZA of 30 degrees. Bold numbers represent the differences of aerosol
613 forcing associated with applying different approaches and “true”; the numbers in parentheses
614 represent the ratio of the absolute difference in aerosol forcing to the results calculated with
615 “true.”

DRF (in Wm^{-2})	Aspen 1	Aspen 2	Aspen 3	Aspen 4	Aspen 5	Aspen 6
True	-7.28	5.79	3.04	6.32	-15.65	-10.37
MEVA	-6.90	6.11	2.58	6.93	-15.09	-9.73
Linear MODIS	-9.88	1.93	0.23	5.31	-16.95	-12.47
Averaged band MODIS	-9.43	4.21	0.44	5.96	-16.99	-12.37
Liang visible and near infrared	-9.18	4.30	1.01	7.89	-17.05	-12.17
Liang shortwave	-1.57	6.98	0.67	7.23	-9.95	-3.49
Differences of DRF (in Wm^{-2})						
MEVA - True	0.38 (5%)	0.32 (6%)	-0.46 (15%)	0.61 (10%)	0.56 (4%)	0.64 (6%)
Linear MODIS - True	-2.60 (36%)	-3.86 (67%)	-2.81 (92%)	-1.01 (16%)	-1.30 (8%)	-2.10 (20%)
Average band MODIS - True	-2.15 (30%)	-1.58 (27%)	-2.60 (86%)	-0.36 (6%)	-1.34 (9%)	-2.00 (19%)
Liang visible and near infrared - True	-1.90 (26%)	-1.49 (26%)	-2.03 (67%)	1.57 (25%)	-1.40 (9%)	-1.80 (17%)
Liang shortwave - True	5.71 (78%)	1.19 (21%)	-2.37 (78%)	0.91 (14%)	5.70 (36%)	6.88 (66%)

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626 Table 6. The calculated daily average aerosol DRF at equator and in equinox (in Wm^{-2} ; from 0.3
627 to 2.35 μm) with different approaches to the surface reflectance spectrum. The last column
628 presents the average aerosol forcing over the six vegetation types. Bold numbers indicate the
629 differences in daily averaged aerosol forcing associated with different approaches; the numbers
630 in parentheses represent the ratio of these differences to the results calculated with “true.” Other
631 parameters used in the simulation include SSA (at 0.55 μm) of 0.89 and AOD (at 0.55 μm) of
632 0.32.

Daily average DRF (in Wm^{-2})	Veg 1	Veg 2	Veg 3	Veg 4	Veg 5	Veg 6	Average over the six vegetation
True	-2.65	-8.68	-5.42	-4.80	-4.94	-5.25	-5.29
MEVA	-2.56	-8.68	-5.51	-4.73	-4.94	-5.04	-5.24
Linear MODIS	-3.91	-9.35	-6.54	-5.94	-6.09	-5.60	-6.24
Averaged band MODIS	-4.03	-9.50	-6.63	-6.14	-6.34	-5.63	-6.38
Liang visible and near infrared	-3.61	-9.57	-6.37	-5.91	-5.78	-4.85	-6.02
Liang shortwave	2.21	-7.32	-2.66	-1.14	0.02	-0.07	-1.49
Difference of daily average DRF (in Wm^{-2})							
MEVA - True	0.09	0.00	-0.09	0.07	0.00	0.21	0.05 (-0.9%)
Linear MODIS – True	-1.26	-0.67	-1.12	-1.14	-1.15	-0.36	-0.95 (18.0%)
Average band MODIS - True	-1.38	-0.81	-1.21	-1.34	-1.40	-0.38	-1.09 (20.6%)
Liang visible and near infrared - True	-0.96	-0.89	-0.95	-1.11	-0.84	0.39	-0.73 (13.8%)
Liang shortwave - True	4.86	1.37	2.76	3.66	4.96	5.17	3.80 (-71.8%)

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640 Figure captions:

641 Fig. 1. Black sky albedo maps at MODIS bands 1-7 calculated from the MODIS file
642 MCD43C1.A2006241.005.2008109074010.hdf with SZA equaling 32 degrees.

643 Fig. 2. The x axis represents the wavelength from 0.3 to 2.35 μm labeled by seven MODIS
644 channels; the dotted red curve is the spectral reflectance for vegetation 5 (*miconia guianensis*)
645 adapted from Arai et al (2010); the green stars represent the corresponding reflectance at MODIS
646 bands 1-7. In this figure, the spectral reflectance results between 1.35 to 1.45 μm and 1.85 to
647 1.95 μm were linearly interpolated based on Arai et al (2010) data. These solid red lines are
648 shown as dots in Fig. 3, 5, 6, and 9.

649 Fig. 3. Traditional approaches to estimate the continuous reflectance spectra based on MODIS
650 bands 1-7: Liang short wave, Liang visible and near infrared, average band MODIS, and linear
651 MODIS. In each subplot, the x axis represents the wavelength from 0.3 to 2.35 μm ; the dotted
652 red curve is the spectral reflectance for *miconia guianensis* as shown in Fig. 2; the green stars
653 represent the corresponding reflectance at MODIS bands 1-7; the solid blue lines represent each
654 traditional approach.

655 Fig. 4. (a) Downward fluxes at the TOA and surface. The x axis is the wavelength (from 0.3 to
656 2.5 μm) labeled with MODIS bands 1-7; the red curve represents the incoming solar radiation at
657 the top of the atmosphere (TOA); the green curve represents the downward radiation reaching
658 the surface. The simulation was done with the following inputs: no boundary layer aerosols,
659 SZA=30 degrees, tropical atmospheric profile, and surface albedo being as the red curve in Fig.
660 2. Fig. 4. (b) The three missing features by linearly connecting the reflectance at MODIS bands
661 1-7. The x axis is the wavelength from 0.3 to 2.5 μm labeled with MODIS bands 1-7; the dotted
662 red curve represents the reflectance as shown in Fig. 2; the solid blue line represents the linearly
663 connected reflectance at MODIS bands 1-7; the shaded areas represent three distinct missing
664 features: missing vegetation red edge feature at around 0.7 μm , and the missing water absorption
665 features at around 1.44 and 1.92 μm .

666 Fig. 5. This figure illustrates four auxiliary channels and the ratios used to determine the MODIS
667 enhanced surface albedo (MEVA), where the x axis represents wavelength (from 0.3 to 2.35 μm)
668 labeled with MODIS bands 1-7 on the bottom and four auxiliary channels (0.69, 0.72, 1.44, and
669 1.92 μm) at the top. Dotted red curve ("True") represents the reflectance spectrum as shown in
670 Fig. 2; the green stars present the corresponding reflectance at MODIS bands 1-7. The
671 reflectance at the four auxiliary channels are determined as: at 0.69 μm , the reflectance is
672 obtained by linearly extrapolating the reflectance at 0.55 and 0.67 μm ; at 0.72 μm , the

673 reflectance is the average of the reflectance at 0.69 and 0.86 μm ; at 1.44 μm , the reflectance is
674 40% of the reflectance at 1.24 μm ; at 1.92 μm , the reflectance is 20% of the reflectance at 1.63
675 μm .

676 Fig. 6. Processes used to develop MEVA. The x axis is wavelength (from 0.3 to 3 μm) labeled
677 with MODIS bands 1-7 on the bottom and the auxiliary channels (0.69, 0.72, the variable
678 channel ending the red edge, 1.44, 1.84, 1.92, and 3 μm) at the top. Four of these auxiliary
679 channels were shown in Fig. 5. For the other three auxiliary channels, the variable channel
680 ending the red edge is the crossing point between the linearly extrapolated line from 0.69 to 0.72
681 μm and the linearly extrapolated line from 1.24 to 0.86 μm ; the auxiliary channel at 1.84 and 3
682 μm are determined by averaging the experimental results of several types of vegetation. The
683 dotted red curve (“True”) represents reflectance spectra as shown in Fig. 2; the green stars
684 present the corresponding reflectance at MODIS bands 1-7; the ovals represent the determined
685 reflectance at auxiliary channels; the solid blue line represents MEVA. The detailed procedures
686 for MEVA are explained in section 2.2.

687 Fig. 7. The procedures for MEVA were applied to the reflectance spectra from the JHU spectral
688 library for dry grass, green grass, conifer, and deciduous. The x axis represents the wavelength
689 from 0.3 to 2.5 μm ; the y axis represents the reflectance; the dotted red curves represent the
690 reflectance spectra from the JHU spectral library; the solid blue lines represent the MEVA
691 results.

692 Fig. 8. In each subplot, the x axis represents wavelength from 0.3 to 2.5 μm ; the y axis represents
693 reflectance. The dotted red curves represent reflectance spectra for aspens from the USGS digital
694 spectral library; the solid blue lines represent the MEVA results.

695 Fig. 9. In each subplot, the x axis represents wavelength from 0.3 to 2.5 μm ; the y axis represents
696 reflectance. The solid blue lines represent the MEVA results; the dotted red curves represent the
697 “true” reflectance spectra (adopted from Arai et al 2010) for the following six Amazonia
698 vegetation types: veg 1: *manilkara Hubert*; veg 2: *couratari guianensis*; veg 3: *lecythis lurida*;
699 veg 4: *genipa Americana*; veg 5: *miconia guianensis*; and veg 6: *litter*.

700 Fig. 10. The plot on the left shows the wavelength dependence of AOD for the aerosol models
701 used in the TOA flux and aerosol forcing calculations; the plots on the right display three
702 different aerosol models used in the flux and aerosol forcing calculations: SSA (at 0.55 μm) =
703 0.95, 0.89, and 0.83.

704 Fig. 11. Difference of the aerosol forcing efficiency (in $\text{Wm}^{-2} \text{AOD}^{-1}$) associated with different
705 approaches to estimate reflectance spectra for vegetation types shown in Fig. 9. In each subplot,
706 three groups indicate the results caused by using three different aerosol models shown in Fig. 10:
707 SSA (at 0.55 μm) = 0.95 (noted as “model 1”), 0.89 (noted as “model 2”), and 0.83 (noted as

708 “model 3”). Other parameters used in the calculations include AOD (at 0.55 μm) = 0.32 and SZA
709 = 30 degrees.

710 Fig. 12. The difference of aerosol forcing efficiency derived from averaging the results shown in
711 Fig. 11 over the six Amazonian vegetation types. Bars represent the standard deviation of the
712 aerosol forcing efficiency differences among the six vegetation types. The labels on the x axis
713 denote different methods to estimate vegetation reflectance spectrum used in the aerosol forcing
714 calculation. The curves in green, blue, and red represent the results for different aerosol models:
715 SSA (at 0.55 μm) = 0.95, 0.89, and 0.83. Other parameters used include AOD (at 0.55 μm) of
716 0.32 and SZA of 30 degrees.

717 Fig. 13. Aerosol forcing when SZA varies from 0 to 90 degrees for vegetation types shown in
718 Fig. 9. Curves in different symbols are associated with different approaches to estimate
719 vegetation reflectance spectrum used in the aerosol forcing calculation. Other parameters used in
720 the calculations include AOD (at 0.55 μm) of 0.32 and SSA (at 0.55 μm) of 0.89.

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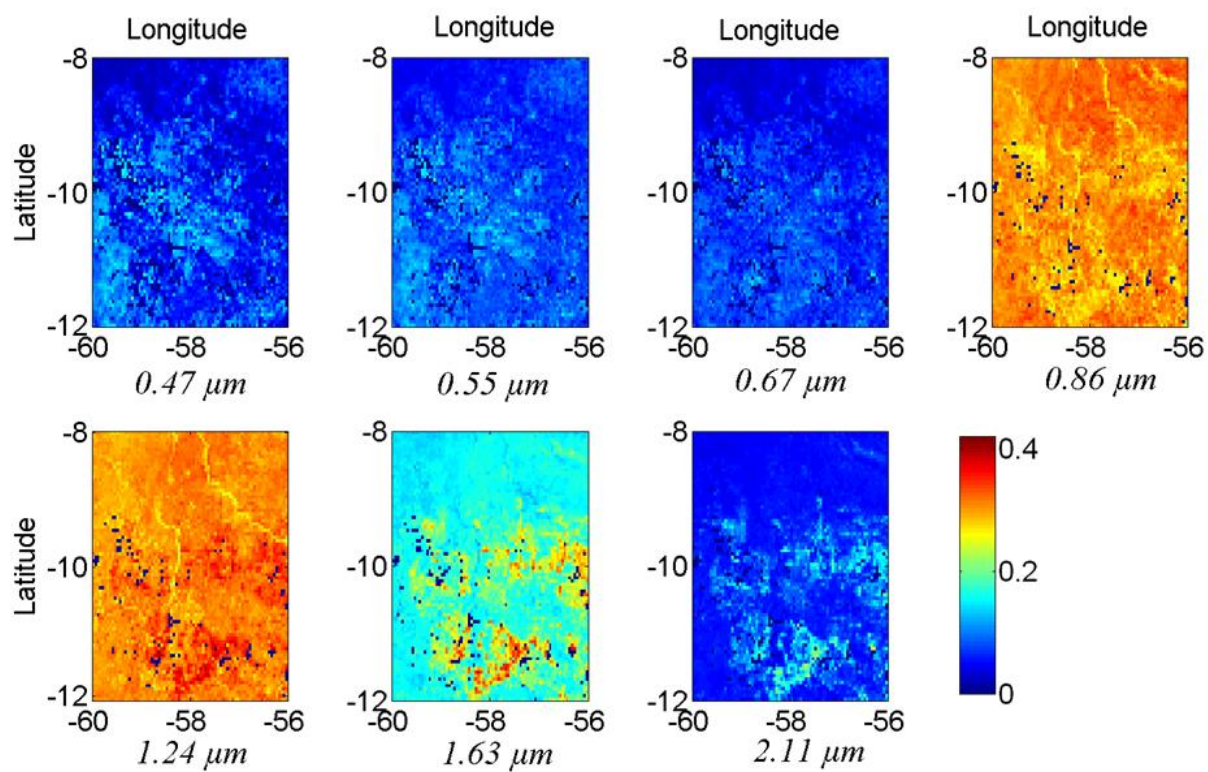
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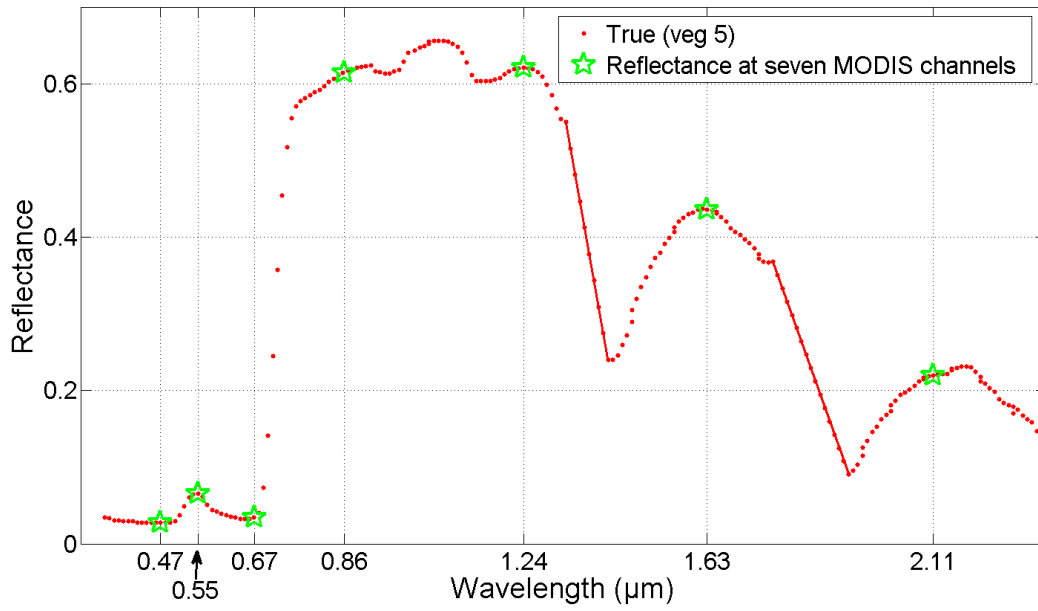
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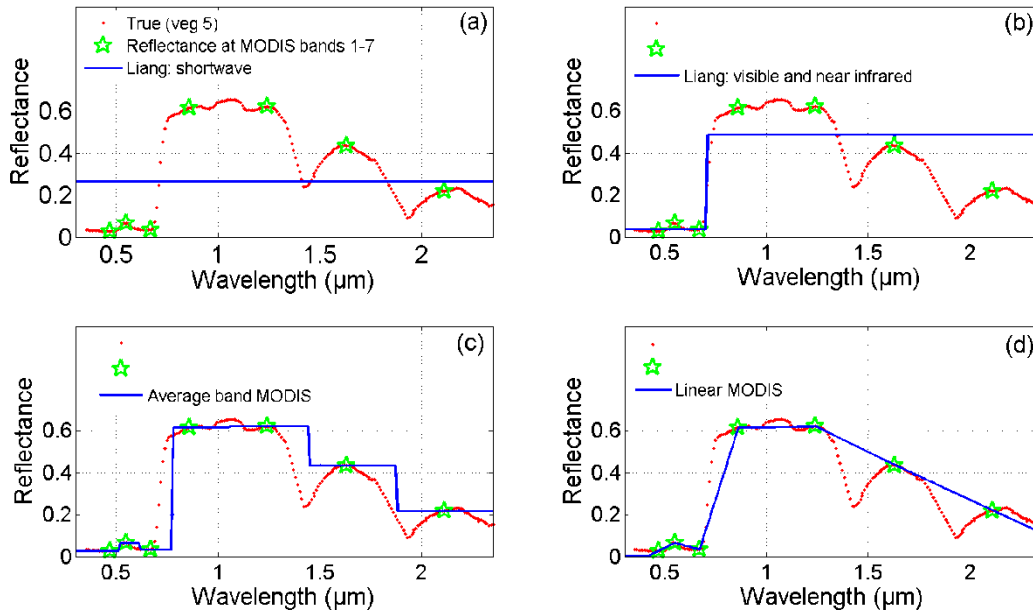
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735 Fig 1



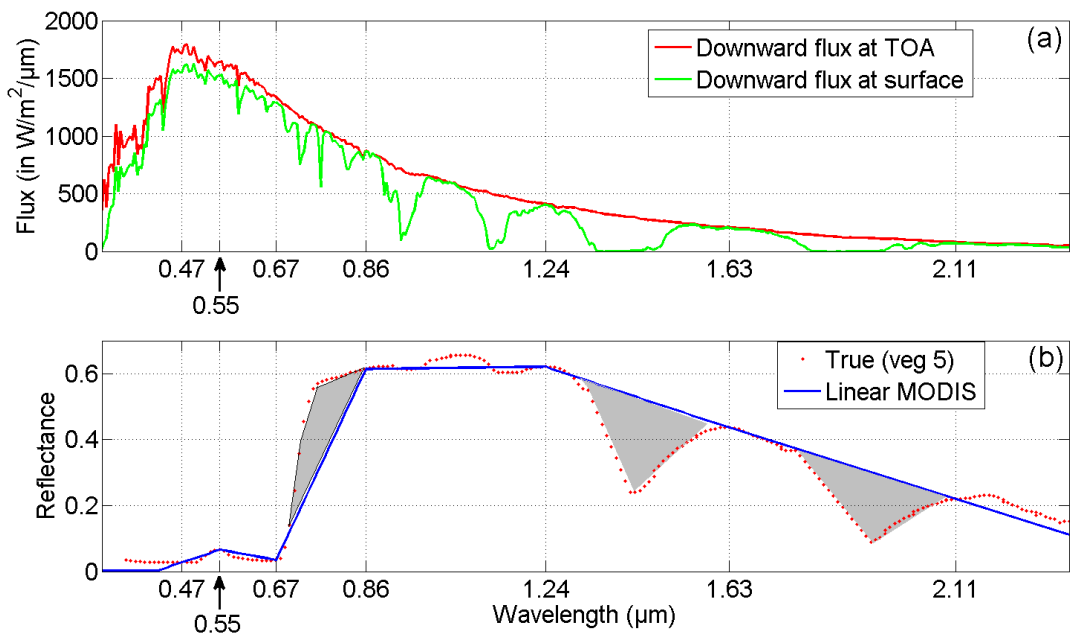
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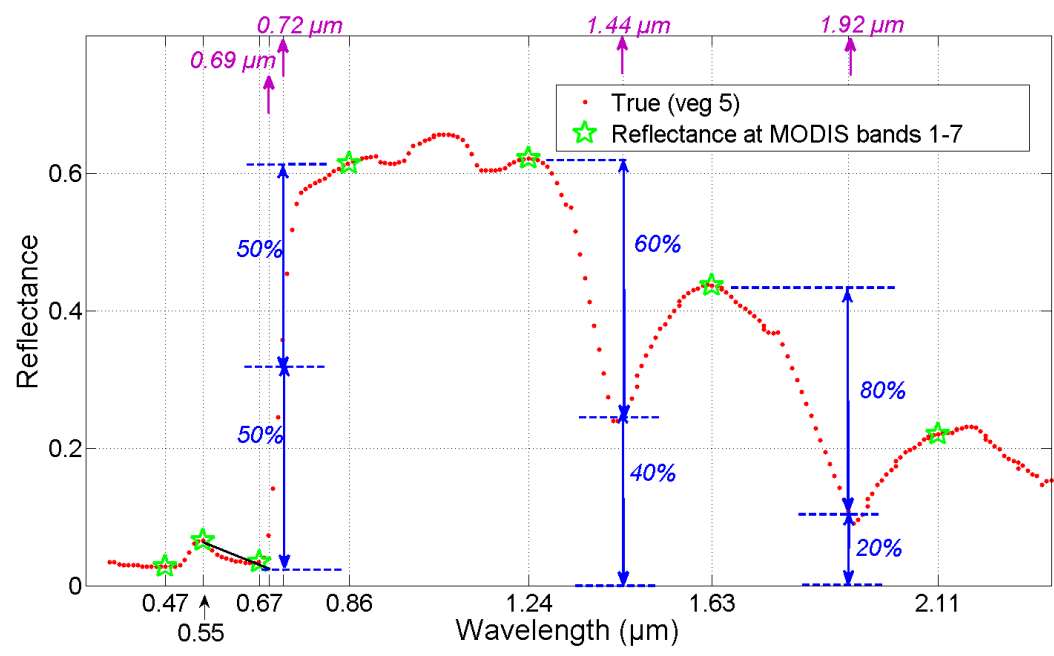
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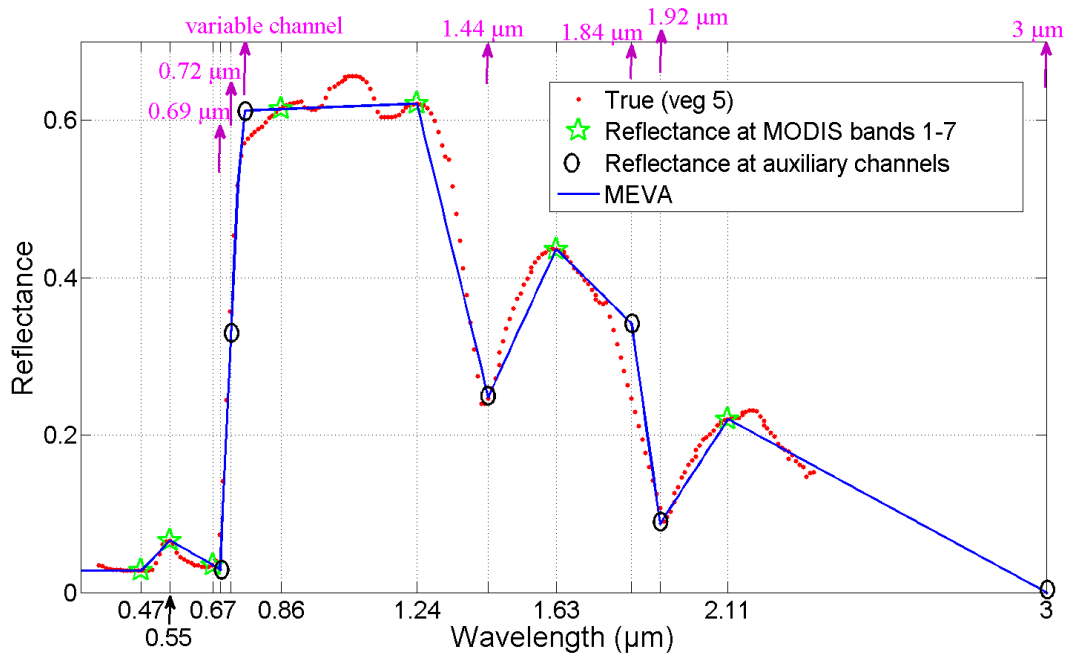
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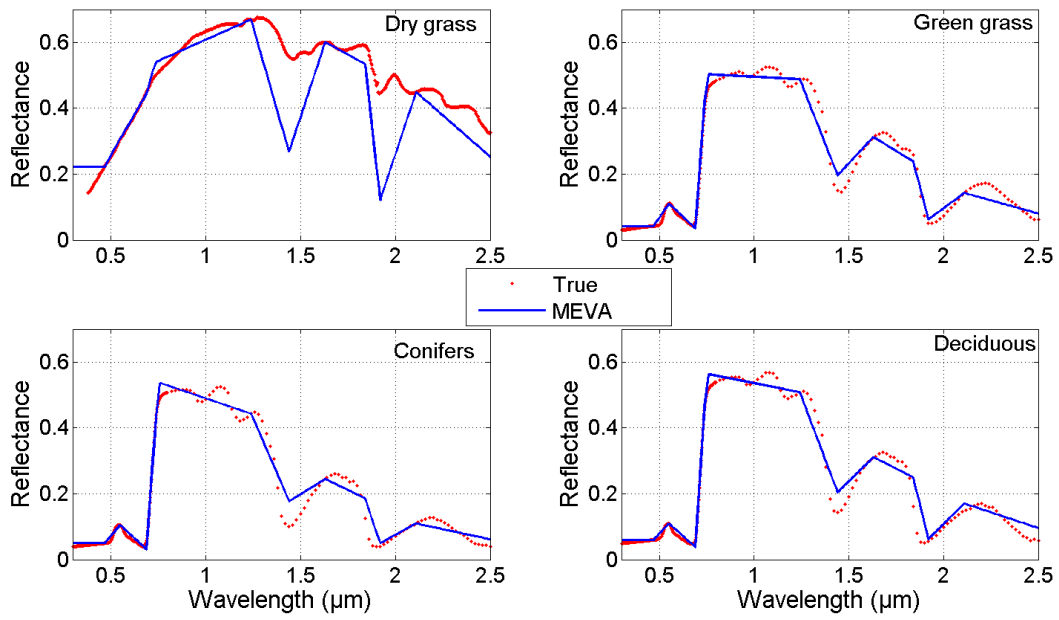
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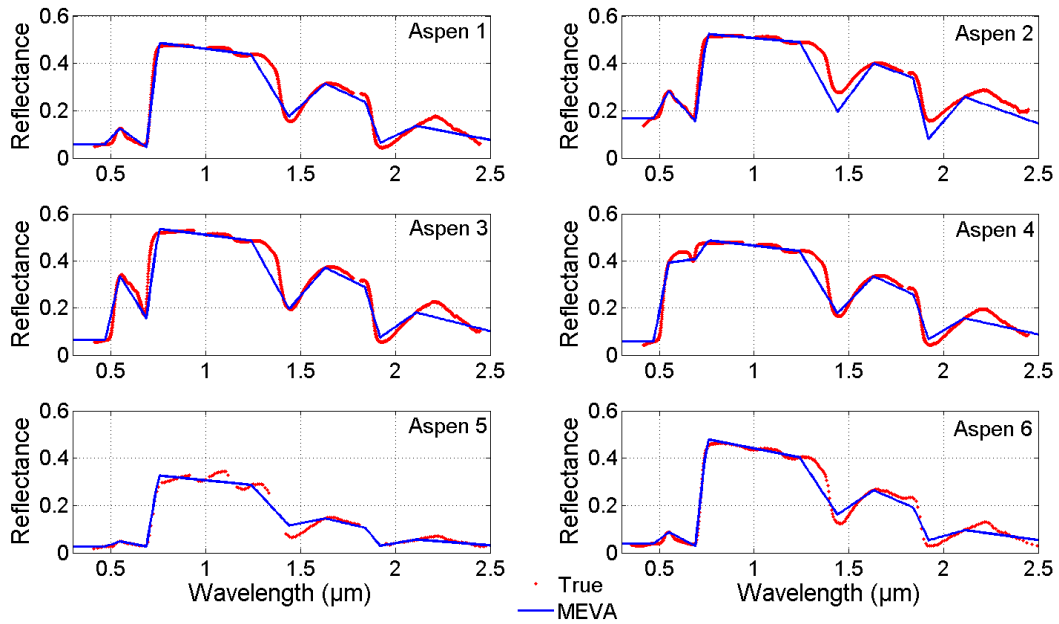
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745 Fig 6

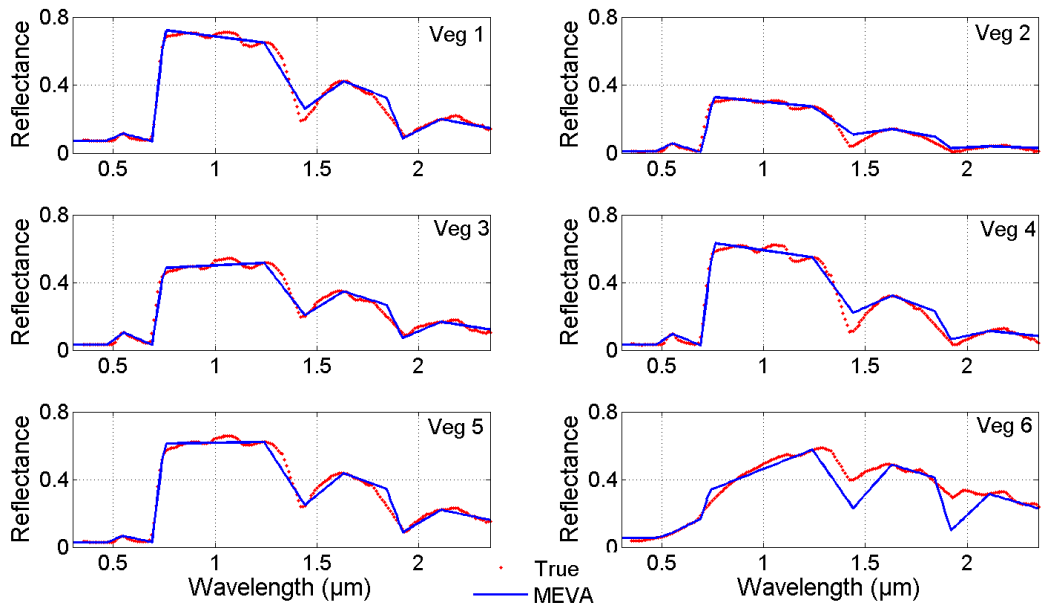


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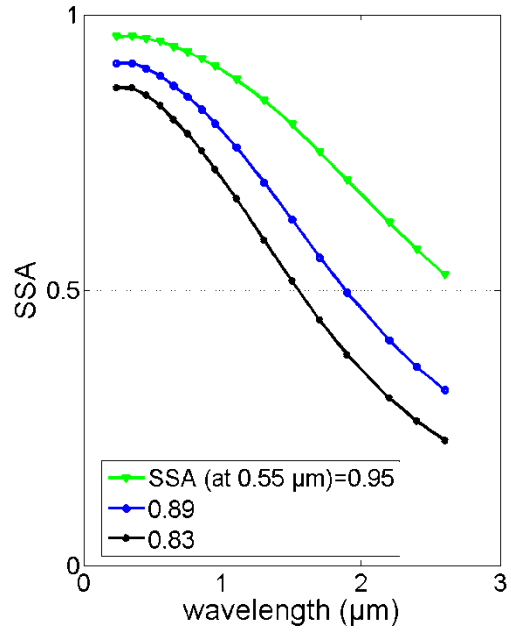
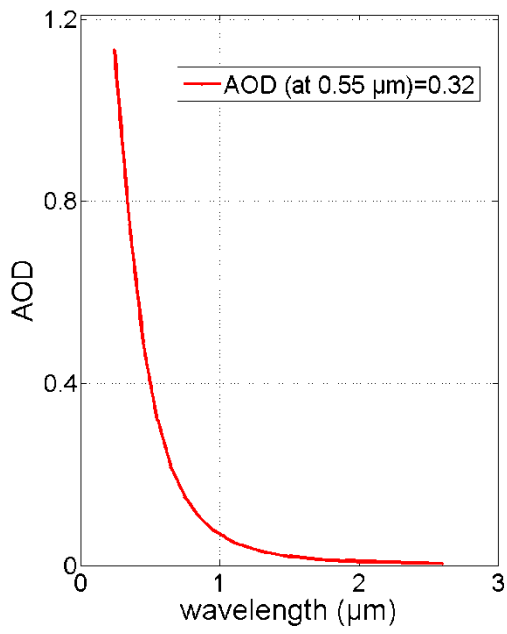
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749 Fig 8



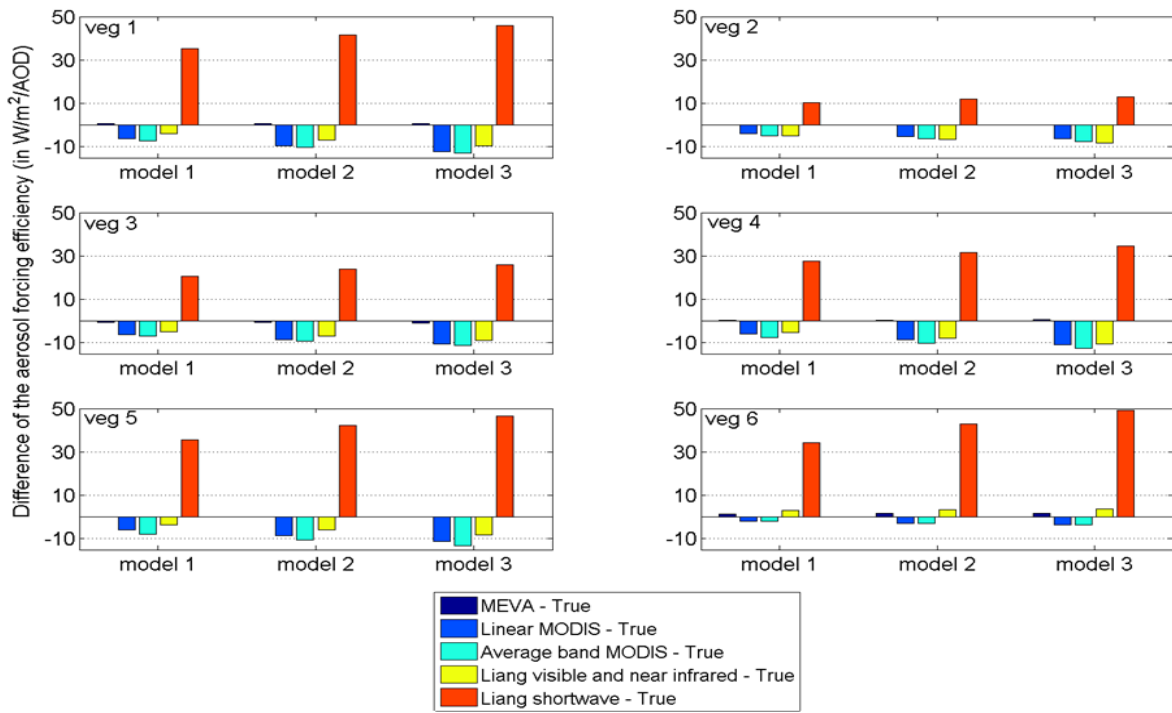
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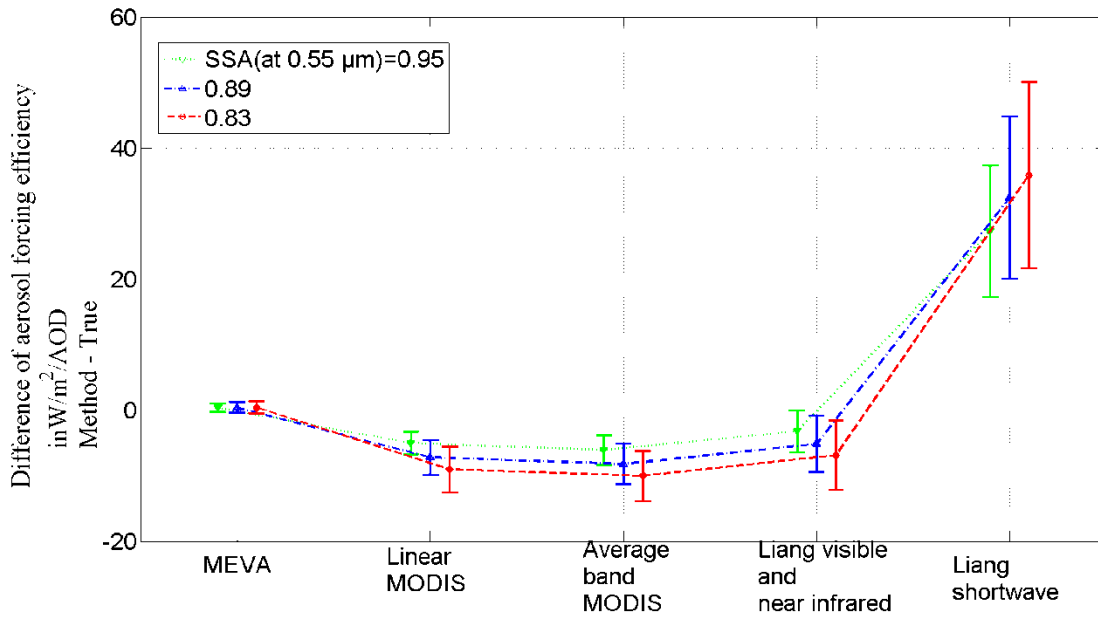
753 Fig 10

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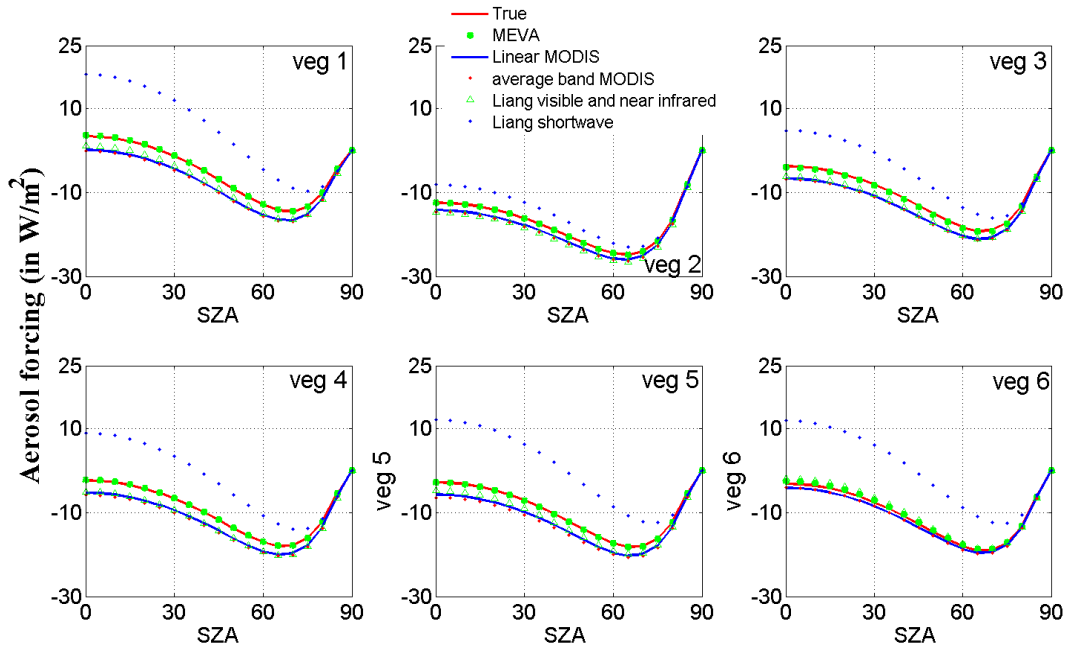
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756 Fig 11



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758 Fig 12



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760 Fig 13