

1 **Comparative Analysis of MODIS, MISR and AERONET Climatology**  
2 **over the Middle East and North Africa**

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10 **Abstract:**

11 Comparative analysis of MISR MODIS, and AERONET AOD products is performed  
12 over seven AERONET stations located in the Middle East and North Africa for the  
13 period of 2000 – 2015. Sites are categorized into dust, biomass burning and mixed.  
14 MISR and MODIS AOD agree during high dust seasons but MODIS tends to  
15 underestimate AOD during low dust seasons. Over dust dominated sites, MODIS/Terra  
16 AOD indicate a negative trend over the time series, while MODIS/Aqua, MISR, and  
17 AERONET depict a positive trend. A deviation between MODIS/Aqua and  
18 MODIS/Terra was observed regardless of the geographic location and data sampling.  
19 The performance of MODIS is similar over the entire region with ~64 percent of AOD  
20 within the  $\Delta\tau = \pm 0.05 \pm 0.15\tau_{AERO}$  confidence range. MISR AOD retrievals fall within  
21 84 percent of the same confidence range for all sites examined here. Both MISR and  
22 MODIS capture aerosol climatology; however few cases were observed where one of  
23 the two sensors better captures the climatology over a certain location or AOD range  
24 than the other sensor. AERONET Level 2.0 Version 3, MODIS Collection 6.1, and  
25 MISR V23 data have been used in analyzing the results presented in this study

26 **Keywords:** AOD; Remote Sensing; North Africa; Middle East; Validation  
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31 **1. Introduction**

32 The Middle East and North Africa host the largest dust source in the world, the Sahara Desert  
33 in North Africa that may be responsible for up to 18 percent of global dust emission (Todd  
34 et al., 2007, Bou Karam et al. 2010, Schepanski et al. 2016). The vast 650,000 km<sup>2</sup> Rub' al  
35 Khali (Empty Quarter) sand desert is a major source of frequent dust outbreaks and severe  
36 dust storms that has major effect on human activity in the Arabian Peninsula (Böer, 1997,  
37 Elagib and Addin 1997, Farahat et al., 2015).

38 Air quality over the Arabian Peninsula has received significant attention during the past 15  
39 years due to unprecedented overall economic growth, and a booming oil and gas industry,  
40 however, air pollution studies are still far from complete. Frequently blowing dust storms  
41 play a significant role in pollutant transport over the Arabian Peninsula; and major  
42 environmental pollution events such as burning of Kuwait oil fields during the 1991, Gulf  
43 War resulted in a large environmental impact on the Arabian Gulf Area (Sadiq and McCain,  
44 1993, and Farahat 2016).

45 Aerosol optical depth, AOD, (also called aerosol optical thickness, AOT) as a parameter  
46 indicates the extinction of a beam of radiation as it passes through a layer of atmosphere that  
47 contains aerosols. Both satellites and ground-based instruments can be used to measure AOD  
48 in the atmosphere, but within the same temporal coordinates and geographic location  
49 different instruments could generate different retrievals (Kahn et al., 2007, Kokhanovsky et  
50 al., 2007, Liu et al., 2008 and Mishchenko et al., 2009).

51 Since the turn of the 21<sup>st</sup> century, an upward trend of remotely sensed and ground-based  
52 AOD and air pollutants was observed over the Middle East and North Africa (El-Askary  
53 2009, Ansmann et al. 2011, Yu et al. 2013, Chin et al. 2014, Yu et al. 2015, Farahat et al.  
54 2016, Solomos et al. 2017). This positive trend is attributed to the increase in the Middle  
55 Eastern dust activity (Hsu et al., 2012) due to changes in wind speed and soil moisture  
56 (Ginoux et al. 2001 and Kim et al. 2013). Yu et al., (2015) concluded that the persistent La

57 Niña conditions (Hoell et al., 2013) have caused increment in Saudi Arabian dust activity  
58 during 2008 – 2012. Energy subsidies also encourages energy overconsumption in the  
59 Middle East and North Africa with little incentive to adopt cleaner technology. Lack of  
60 applying strict environmental regulations have permitted exacerbated urban air pollution.  
61 During the last two decades, a large number of satellites, ground stations and computational  
62 models contributed to build global and regional maps for the temporal and spatial aerosol  
63 distributions. While, ground-based stations and field measurements can identify aerosols  
64 properties over specific geographic locations, the sparse and non-continues data from ground-  
65 based sensors scattered over the Middle East and North Africa is not sufficient to provide  
66 information on spatial and temporal trends of particulate pollution. On the other hand,  
67 satellites imagery could provide a significant source of data mapping over larger areas.  
68 For its wide spatial and temporal data availability space-born sensors are important sources  
69 to understand aerosols characteristics and transport, however low sensitivity to particle type  
70 under some physical conditions, high surface reflectivity, persistent cloud, and generally low  
71 aerosol optical depth could limit satellite data application in characterizing properties of  
72 airborne particles, especially in the Middle East.

73 In order to evaluate the efficiency of space-borne sensors in representing ground observations  
74 recorded by AERONET stations we have performed detailed statistical inter-comparison analysis  
75 between satellite AOD products and AERONET for seven stations in the Middle East and North  
76 Africa representative for dust, biomass burning, and mixed aerosol conditions (Dubovik et al.,  
77 (2000, 2002, 2006), Holben et al. (2001), Derimian et al., (2006), Basart et al. (2009), Eck  
78 el. (2010), Marey et al., 2010, Abdi et al., (2012)). Previously we analysed these seven  
79 AERONET stations to understand particles categorization and absorption properties (Farahat  
80 et al. 2016), and the current study extends the analysis to the satellite datasets.

81 In the first part of this article, we validated MISR and MODIS retrievals against collocated  
82 AERONET observations. We also assessed the consistency in aerosol trends between space-  
83 borne sensors and ground-based data.

84 In the second part, we evaluated representativeness of satellite-derived aerosol climatology  
85 over the study region from the long-term AERONET data for MISR and MODIS AOD  
86 products. It is especially relevant for the MISR instrument, as its sampling is limited by once  
87 per week observations of the same region from the two overlapping paths. MODIS provides  
88 nearly daily observations to the same geographic location; however, the quality of the product  
89 diminishes over the bright targets potentially affecting MODIS-derived aerosol climatology.  
90 The collocated MISR, MODIS and AERONET data were obtained at the MAPSS website  
91 (<http://giovanni.gsfc.nasa.gov/mapss.html>).

92

## 93 **2. Materials and Methods**

### 94 **2.1 MISR**

95 The Multi-angle Imaging SpectroRadiometer (MISR) instrument to measure tropospheric  
96 aerosol characteristics through the acquisition of global multi-angle imagery on the daylight  
97 side of Earth. MISR applies nine Charge Coupled Devices (CCDs), each with 4 independent  
98 line arrays positioned at nine view angles spread out at nadir, 26.1°, 45.6°, 60.0°, and 70.5°.  
99 In each of the nine MISR cameras, images are obtained from reflected and scattered sunlight  
100 in 4 bands blue, green, red, and near-infrared with a centre wavelength value of 446, 558,  
101 672, and 867 nm respectively. The combination of viewing cameras and spectral wavelengths  
102 enables MISR to retrieve aerosols AOD over high reflection surfaces like deserts.

103 In this study, we use Level 2 (ver. 0023) AOD at 558 nm (green band) measured by MISR  
104 instrument with a 17.6 km resolution aboard the Terra satellite. MISR Level 2 aerosol  
105 retrievals use only data that pass angle-to-angle smoothness and spatial correlation tests

106 (Martonchik et al. 2002), as well as stereoscopically derived cloud masks and adaptive cloud-  
107 screening brightness thresholds (Zhao and Di Girolamo, 2004).

## 108 **2.2 MODIS**

109 The Moderate Resolution Imaging Spectroradiometer (MODIS) is a payload instrument on  
110 board the Terra and Aqua satellites. Terra and Aqua orbit around the Earth from North to  
111 South and South to North across the equator during the morning and afternoon respectively  
112 (Kaufman et al., 1997). Terra MODIS and Aqua MODIS provides nearly daily coverage of  
113 the Earth's surface and atmosphere in 36 wavelength bands, ranging from 0.412 to 41.2  $\mu\text{m}$ ,  
114 with spatial resolutions of 250 m (bands 1-2), 500 m (bands 3-7), 1000 m (bands 8-36).  
115 Located near-polar orbit (705 km), MODIS has swath dimensions of 2330 km  $\times$  10 km and  
116 a scan rate of 20.3 rpm. With its high radiometric sensitivity and swath resolution MODIS  
117 retrievals provide information about aerosols optical and physical characteristics. MODIS  
118 uses 14 spectral band radiance values to evaluate atmospheric contamination and determine  
119 whether scenes are affected by cloud shadow (Ackerman et al., 1998).

120 The Deep Blue (DB) is a NASA developed algorithm to calculate AOD over land using  
121 MODIS data. By measuring contrast between aerosols and surface features, DB retrieves  
122 AOD. Over bright land, Deep Blue uses (0.412, 0.470/0.479  $\mu\text{m}$ ) for AOD retrievals. Over  
123 water, the DB algorithm is not used, but the Dark Target (DT) algorithm is used instead.

124 The MODIS DT algorithm is designed for aerosol retrieval from MODIS observations, over  
125 dark land surfaces (low values of surface reflectance) (e.g., dark soil and vegetated regions)  
126 in parts of the visible (VIS, 0.47 and 0.65  $\mu\text{m}$ ) and shortwave infrared (SWIR, 2.1  $\mu\text{m}$ )  
127 spectrum (Kaufman et al., 1997). Level 2 – Collection 6 (C006) of the algorithm are used to  
128 retrieve MODIS aerosols' time series data. Levy *et al.* (2010) reported that the dark-target  
129 algorithm AOD at 550 nm measurement for Collection 5 (C005) includes uncertainty of  $\pm$   
130  $(0.05\tau+0.03)$  and  $\pm (0.15\tau+0.05)$  over ocean and land, respectively. This uncertainty is

131 caused by uncertainties in computing cloud masking, surface reflectance, aerosol model type  
132 (e.g., single scattering albedo), pixels selections and instrument calibration.  
133 Both DB and DT algorithms have been used in this study. DB data were used over land,  
134 while DT retrievals were used over water. For regions like Bahrain where large water body  
135 surrounds land, a combined DB and DT algorithm for land and ocean has been used. This is  
136 because the MODIS matched ground-based AERONET station in Bahrain (described in  
137 section 2.3 and Table 1) is located less than 2 km from the coastline. This makes MODIS  
138 combine retrievals for both land and water over this region. Data are available at  
139 <https://giovanni.gsfc.nasa.gov/giovanni>.

### 140 **2.3 AERONET**

141 The Aerosol Robotic Network (AERONET) (Holben et al., 1998 and Holben et al., 2001) is  
142 a ground-based remote sensing aerosols network that provides a long-term data related to  
143 aerosol optical, microphysical and radiative properties. With over 700 global stations, the  
144 AERONET data is widely used in validating satellite retrievals (Chu et al., 1998 and  
145 Higurashi et al., 2000).

146 The sun photometers used by AERONET measure spectral direct-beam solar radiation, as  
147 well as directional diffuse radiation in the solar almucantar. The former are used to determine  
148 columnar spectral AOD and water vapour, provided at a temporal resolution of  
149 approximately 10–15 min (Sayer et al. 2014). AERONET direct-sun AOD has a typical  
150 uncertainty of 0.01–0.02 (Holben et al., 1998) and is provided at multiple wavelengths at  
151 340, 380, 440, 500, 675, 950, and 1020 nm.

152 Seven AERONET sites were selected for satellite validation in this study (Table 1.). The sites  
153 were selected based on their geographic locations to represent aerosols characteristics over  
154 North Africa and the Middle East (Farahat et al., 2016). A record of long-term data collection  
155 was another factor in the selection process.

156 **Data Matching Approach**

157 Multi-sensors data matching requires using only compatible data to eliminate uncertainties  
158 associated with cloud shadow and spatial and temporal retrievals produced by different  
159 instruments (Liu and Mishchenko (2008) and Mishchenko et al., 2009).

160 The comparison of MISR and MODIS products against AERONET is performed to evaluate  
161 satellites' retrieval over individual North Africa and Middle East sites (see Table 1). There  
162 is only a small number of AERONET measurements that are perfectly collocated with  
163 MODIS and MISR. One way to work with this lack of compatibility problem is to compare  
164 satellites measurements nearby a certain AERONET site and comparing AERONET  
165 measurements nearly synchronized with the satellite overpass time (Sioris et al. 2017).  
166 Another reasonable strategy is to average all satellite measurements with a certain distance  
167 of an AERONET location and average all AERONET measurements within a certain time  
168 range (Mishchenko et al., 2010). The results presented in this paper are based on the second  
169 approach as it compares average spatial satellite measurements with average temporal  
170 AERONET measurements. We implemented the Basart et al., (2009) approach in using a  
171 spatial and temporal threshold of 50 km and 30 min for MISR, MODIS, and AERONET data  
172 matching.

173 We use the Giovanni Multi-sensor Aerosol Products Sampling System MAPSS  
174 (<http://giovanni.gsfc.nasa.gov/aerostat/>) for the data inter-comparison as aerosols products  
175 are averaged from measurements that are within a radius of ~ 27.5 km from the AERONET  
176 station and within 30 min of each satellite flyover over this location. These data are  
177 represented in the article by MISR / MODIS “matched AERONET data”.

178 “All data” represents AOD products at the selected station. AERONET station ‘all data’  
179 are obtained through AEROSOL ROBOTIC NETWORK (AERONET) website  
180 (<https://aeronet.gsfc.nasa.gov/>). Daily AOD data with level 2.0 quality was used in the

181 analysis (Smirnov et al., 2000) . Level 2.0 AOD retrievals are accurate up to 0.02 for mid-  
182 visible wavelengths.

183 MISR ‘all data’ is available through MISR website (<https://www.misr.jpl.nasa.gov>).

184

### 185 **3. Statistics**

186 We have used two statistical parameters to compare data retrievals from space-borne and  
187 ground based sensors including:

188 (1) Correlation coefficient (R),

189 The correlation coefficient is a parameter to measure data dependence. If the value of R is  
190 close to zero, it indicates weak data agreement. And values close to 1 or -1 indicate that data  
191 retrievals are positively or negatively linearly related (Cheng et al., 2012).

192

193 (2) Good Fraction (G- fraction).

194 The G- fraction indicator uses a data confidence range defined by MISR and MODIS  
195 (Bruegge et al., 1998 and Remer et al., 2005) over the land and ocean that combines absolute  
196 and relative criterion and weights data equally such that small abnormalities will not affect  
197 the inter-comparison statistics (Kahn et al., 2009). In this study, we use MODIS confidence  
198 range which defines data retrieval as “good” if the difference between MODIS and  
199 AERONET is less than

$$200 \quad \Delta\tau = \pm 0.03 \pm 0.05\tau_{AER}, \quad \text{Over ocean,} \quad (1)$$

$$201 \quad \Delta\tau = \pm 0.05 \pm 0.15\tau_{AER}, \quad \text{Over land.} \quad (2)$$

202

203 where  $\tau_{AER}$  is the optical depth retrieved using AERONET stations. The G-fraction is the  
204 percentage of MODIS data retrievals that satisfies (Equations (1) and (2)) over ocean and



205 land respectively. Optical depth threshold over land (Equation (1)) is higher than over ocean  
206 (Equation (2)) due to harder data retrievals and high data instability over land.

207 An advantage of using data confidence range is excluding small fraction data outliers from  
208 producing inexplicably large influence on comparison statistics by weighting all events  
209 equally.

210

## 211 **4. Results and discussion**

### 212 **4.1 Validating MISR and MODIS AOD retrievals against AERONET observations** 213 **over the Middle East and North Africa**

214 Illustrated in Figures 2, 3 and Tables 2, 3 is a regression analysis of MISR and MODIS Terra  
215 AOD products against AERONET AOD over the seven AERONET sites, shown in Table1,  
216 from 2000 – 2015.

217 The correlation coefficient between MISR and AERONET AOD at region 1 is equal to or  
218 above 0.85 except in Bahrain during DJF and JJA (Figure (2) and Table 2), which could be  
219 attributed to lack of data and the impact of water surface reflectivity over Bahrain. Similar  
220 correlation coefficient values were found in region 2 where MISR-AERONET AOD shows  
221 smaller deviations from the MODIS data (Figures (2, 3) and Table 3). In general, MODIS-  
222 AERONET AOD correlation coefficient is lower than those of MISR at all sites, except  
223 Mezaira, where MISR and MODIS matched AERONET AOD correlation almost match.  
224 The lowest MODIS-AERONET AOD correlation coefficient was found over Cairo but could  
225 be attributed to the lack of data availability at this location (Figs 3e-h). Low values of  
226 MODIS-AERONET correlation coefficient is also found over Saada, Taman, and Sedee  
227 Boker sites.

228 Over all AERONET stations, the number of MODIS AERONET matched AOD are 4 to 8  
229 times those of MISR which is expected from the MISR's sampling.

230 Comparisons show that the difference between MISR and MODIS retrievals at the selected  
231 AERONET sites could be significant as expected from the MODIS DT algorithm  
232 performance over bright land surfaces Kokhanovsky et al. (2007).

233 High AOD values over regions 1 and 2 measured by both AERONET and satellites' sensors  
234 indicate higher dust activities that peaks during May – Aug during dust storms season. Higher  
235 AOD values recorded during SON over Cairo station could be caused by seasonal rice straw  
236 burning by farmers in Cairo, an environmental phenomena known as Cairo Black cloud  
237 (Marey et al. 2010). As shown in (Figure (3)), the daily variability in MODIS measurements  
238 is larger than that of MISR in all the three regions. In general, MODIS tends to underestimate  
239 the AOD values on low dust seasons (Figures (2, 3) and Tables 2, 3).

240 The MODIS underestimated AOD values are more noticeable over Bahrain. This could be  
241 attributed to large water body surrounding Bahrain, which should affect surface reflectivity.  
242 Moreover, water in the Arabian Gulf has been polluted in recent years (Afnan 2013), leading  
243 to possible changes in watercolour and uncertainties in calculating surface reflectivity. The  
244 patchy land surface or pixel grid contaminated by water body is the dominant error sources  
245 for MODIS aerosol inversion over the land areas (He et al. 2010).

246 Compared to MODIS, MISR's outperform in retrieving AOD over region 1 including vast  
247 highly reflecting desert areas can be attributed to its multispectral and multi-angular  
248 coverage, which make MISR provide better viewing over a variety of landscapes.  
249 Meanwhile, MISR retrieval also takes into consideration aerosols' particles nonsphericity,  
250 which could have significant effect on its AOD retrievals (von Hoyningen-Huen and Posse  
251 1997). MISR's retrieval did not perform well over Cairo site due to lack of matched points  
252 in most of the seasons (15 in DJF, 39 in MAM, 61 in JJA, and 23 in SON during 2000 -  
253 2015).

254

255 **4.2 Trends of AOD MISR, MODIS, and AERONET retrievals over the Middle East**  
256 **and North Africa**

257 Figure 4 shows time series of monthly mean AOD derived from MODIS/Aqua,  
258 MODIS/Terra, MISR and AERONET over a) dust b) biomass and c) mixed dominated  
259 aerosol regions. The satellite AOD trends are calculated from the data collocated with  
260 AERONET observations.

261 MODIS/ Aqua and MISR AOD at Solar Village have positive trends, while MODIS/ Terra  
262 AOD have negative trends along time series (Fig. 4a). MODIS-Aqua AOD differ from those  
263 of MODIS-Terra. Discrepancy between Aqua and Terra retrievals could be related to  
264 instrument calibration, or the difference in aerosol and cloud conditions from the morning to  
265 the afternoon. Both MODIS Aqua and Terra are underestimating AOD at Solar Village.

266 MISR AOD trend shows a better agreement with Solar Village AERONET AOD as  
267 compared to MODIS.

268 Both MODIS/Aqua and MODIS/Terra AOD show a stable trend over time at Mezaria site  
269 (not shown in the figure) with a correlation coefficient of 0.11 and 0.04 respectively.

270 MODIS/Aqua AOD over Bahrain (not shown in the figure) show, less time trend stability  
271 compared to those at Solar Village with a correlation coefficient 0.63. MODIS/Aqua,

272 MODIS/Terra, and MISR AOD depicts a positive trend over Cairo (Fig. 4b). Taman site  
273 (Fig. 4c): MODIS/Aqua, MODIS/ Terra, MISR AOD agrees with Taman AERONET on a  
274 positive trend indicating data stability over this site.

275 Long-range (2000 – 2015) tendency indicates that contradictory AOD trend of Terra and  
276 Aqua is site-dependent and does not necessarily apply everywhere.

277 AOD difference between Terra and Aqua could be used as another indicator of the long-term  
278 satellites performance. AOD difference (Terra AOD minus Aqua AOD) varies from -0.01 to  
279 0.19, -0.10 to 0.18 over Solar Village and Taman respectively (Fig. 5). Over the Solar

280 Village, Terra overestimates AOD during 2002-2004 and underestimates the AOD after  
281 2005. Taman shows similar trend, however over/underestimation amount is not unique for  
282 all sites. This is an indication that Aqua and Terra retrievals disagreement takes place  
283 regardless of the region but site sampling has significant effect on the amount of  
284 contradiction.

285 Statistical comparison between MISR and MODIS/Terra AOD at corresponding AERONET  
286 stations is performed by calculating G-fraction using  $\Delta\tau = \pm 0.05 \pm 0.15\tau_{AERO}$  as a confidence  
287 interval. Over the region 1, MISR AOD retrievals are more accurate than MODIS retrievals.  
288 MODIS, however, performs better over region 2 sites with high percentage of the data points  
289 falling within the confidence range (Tables 2 and 3). High light reflections from the desert  
290 landscape surrounding region 1 could have an effect on MODIS retrievals.

291 Excluding Bahrain and Cairo for low data retrievals the performance of MODIS tends to be  
292 similar over all region with ~ 64 percent of AOD retrievals fall within the  
293  $\Delta\tau = \pm 0.05 \pm 0.15\tau_{AERO}$  confidence range of the AERONET AOD while MISR retrievals  
294 show better performance with ~ 84 percent of the data falling within the same confidence  
295 range. This could be attributed to low number of retrievals available for Bahrain and Cairo  
296 compared to other sites. Vast sea region surrounding Bahrain and complex landscape in Cairo  
297 could also have an impact on retrievals.

### 298 **4.3 Evaluating the MISR and MODIS climatology over Middle East and North Africa**

299 Comparisons between MISR and MODIS AOD at selected AERONET stations over the  
300 2000 – 2015 period are shown in Figures 6- 12.

301 Figure (6a, b) shows histogram of the MISR, MODIS and AERONET AOD at Solar Village  
302 for MISR and MODIS data points collocated with AERONET observations. The mean,  
303 standard deviation, and number of measurements are also presented.

304 MISR tends to underestimate the frequency of low AOD compared to AERONET but  
305 overestimate the frequency of high AOD. MISR histograms show prominent peaks at 0.50  
306 that can be also observed in AERONET and at 0.75 that could not be seen in AERONET.  
307 MISR and AERONET AOD climatology agree well with one another. MODIS also tends to  
308 underestimate the frequency of low AOD events and overestimate the frequency of high  
309 AOD events. High surface reflectance could cause overestimation in MODIS AOD (Ichoku  
310 et al., 2005). Both MISR and MODIS provide a good representation of the AOD climatology  
311 as compared to AERONET at the Solar Village. Mezaria station, which is located in an arid  
312 region in the UAE, has a similar climatology to the Solar Village site with dust dominating  
313 aerosol. Figure (7a, b) shows histograms of the MISR, MODIS and AERONET AOD at  
314 Mezaria.

315 Similar to the Solar Village, there is a big difference between the number of samples in the  
316 matched data set and full AERONET climatology. For MISR there are 213 matched cases  
317 and for MODIS there are 498 compared to the 2245 for the entire site. This has an impact on  
318 the overall assessment showing significant differences between the matched data and the full  
319 climatology for both MISR and MODIS. First, for the MISR case, the matched AERONET  
320 data have the highest frequency at AOD of 0.15 and 0.35, but the climatology shows the  
321 highest frequency at an AOD of 0.25. AOD in the range of 0.25 to 0.30 are undersampled  
322 relative to the climatology, and AOD more than 0.35 matches the climatology with less than  
323 2 percent AOD greater than 0.85. MODIS matched AERONET data show prominent peaks  
324 at 0.3 and 0.4 compared to the climatology that has a single peak at 0.30.

325 For AOD values between 0.25 and 0.40 MODIS data were found to be under-sampled similar  
326 to MISR data between 0.65 to 0.70 and at 0.35.

327 MISR AOD retrievals matched to AERONET capture the variability in the distribution, but  
328 as in the case of Solar Village the frequency of low AOD events is underestimated but the

329 frequency of high AOD events matched AERONET data. MISR also captures events with  
330 AOD greater than 1. A similar situation is seen in the MODIS comparison, but MODIS  
331 appears to do a better job capturing the overall shape of the AERONET AOD histogram for  
332 this site.

333 The Bahrain AERONET site is located in Manama fairly close to the Arabian Gulf, a location  
334 very different from the previous two sites. The site is also located in an urban area suffers  
335 from significant load of anthropogenic aerosols as a consequence of rapid aluminium  
336 industrial development (Farahat 2016). Figure (8a, b) shows histogram of the MISR, MODIS  
337 and Bahrian AERONET measurements with statistical analysis displayed. The AERONET  
338 data matched to MISR show significant peaks at 0.20, 0.30, 0.45, 0.55, 0.7, 0.8, and 0.95 not  
339 seen in the all data climatology that has peaks at 0.55 and 0.70. AOD less than 0.15 are not  
340 representative in the matched data set at all. MISR is representing the peaks at 0.45 in the  
341 matched data set but misses the peaks at 0.20, 0.30, and 0.35. The MISR climatology agrees  
342 well with the AERONET all data climatology for all AOD. MODIS on the other hand shows  
343 an extremely large frequency of AOD at 0.1 not represented by AERONET coupled with an  
344 underestimation of AOD greater than 0.3. This could be attributed to the size of the matching  
345 window and MODIS retrievals preferentially coming from the Arabian Gulf.

346 SAADA station is located close to some hiking trails at the Agoundis Valley in the Atlas  
347 Mountains about 197 km from the city of Marrakesh.

348 MISR AOD matched to AERONET agree well with MISR full climatology retrievals over  
349 SAADA station. Both retrievals slightly underestimate SAADA full climatology and over  
350 estimate SAADA matched data retrievals at AOD equal to 0.2 while show good agreement  
351 for AOD greater than 0.2. MODIS matched to AERONET retrievals overestimate the  
352 frequency of AOD greater than 0.3. While MODIS AOD matched to AERONET captures  
353 climatology at AOD between 0.2 to 0.25, AOD frequency retrievals are under-sampled at

354 AOD between 0.1 to 0.15 with about 13 % less events than SAADA all data retrievals at  
355 AOD equal to 0.1.

356 Figure (9a, b) indicates right skewed distribution of SAADA AOD towards small AOD  
357 values with 10.3 % and 30.1 % of AOD > 0.4 as measured by MISR and MODIS  
358 respectively. Taking into consideration MODIS overestimation we conclude that SAADA  
359 site is characterized by small AOD values and this could be related to the land topology  
360 where the station is located.

361 While MISR is capturing high AOD climatology over SAADA, both MISR and MODIS  
362 are underestimating the frequency of lower AOD events. Nevertheless, MISR captures the  
363 climatology of AOD less than 0.1 missed by MODIS retrievals.

364 Taman AERONET station is located at the oasis city of Tamanrasset, which lies in Ahaggar  
365 National Park in southern Algeria.

366 Figure (10 a, b) depicts that Taman AERONET AOD climatology is similar to those at  
367 SAADA and has a high frequency of low AOD events. Both MISR AOD matched to  
368 AERONET and MISR all data do not well capture the frequency of AOD less than 0.1 or  
369 larger than 1 while well describe the climatology for AOD in the range of 0.1 to 1. MODIS  
370 AOD matched data to AERONET correctly describe climatology with slight overestimation  
371 of AOD frequencies between 0.05 – 0.15 while not capturing AOD frequencies greater than  
372 1. MISR and MODIS show similar prominent peaks at 0.1 and 0.25 not observed in Taman  
373 AERONET AOD climatology, with more peaks observed by MISR at 0.5, 0.75, and 0.85.

374 Average AOD in SAADA and Taman is ~ 50 percent less than observed at Solar Village,  
375 Mezaria, and Bahrain sites.

376 Except for AOD greater than 1 where ground observations could be more robust, both MISR  
377 and MODIS retrievals can provide very good climatology matching over Taman site.

378 Taking into consideration lower number of MISR matching AERONET observations  
379 compared to MODIS ~ 21 and 49 percent over SAADA and Taman respectively, MISR is  
380 outperforming over these two sites, which can be attributed to its multiangle viewing  
381 capabilities over complex terrains including mountainous areas (Atlas Mountains).

382 Cairo is a mega city well known for its high pollution due to traffic and agriculture activities.  
383 MISR and MODIS matched data correctly capture AOD climatology over Cairo compared  
384 to AERONET as shown in Figure (11a, b). MISR retrievals collocated with AERONET over  
385 estimate prominent peaks of AERONET AOD at 0.15 – 0.35 while underestimate  
386 AERONET AOD greater than 0.35. MISR ‘all data’ AOD climatology over Cairo station  
387 agrees better with AERONET AOD climatology vs. collocated dataset with some  
388 oversampling at 0.25. Frequency of high AOD retrievals greater than 0.8 have not been  
389 captured by MISR matched or all data retrievals. MODIS matched to AERONET AOD are  
390 also able to well represent Cairo climatology data with a high overestimation of AOD  
391 frequency between 0.05 - 0.2 and an underestimation of AOD larger than 0.4.

392 The complex landscape and local emissions in Cairo could impose major challenges in  
393 MODIS AOD retrievals. Moreover, Cairo is one of the most densely populated cities in the  
394 world that hosts major commercial and industrial centers in North Africa. Cairo also has  
395 complicated aerosols structure developed by long range transported dust in the spring,  
396 biomass burning in the fall, strong traffic and industrial emissions (Marey et al., 2010).

397 Over Cairo station, MODIS correctly represents ground observations for AOD between 0.2  
398 - 0.4 while MISR all data better represents AOD climatology for AOD greater than 0.4.

399

400 MISR, MODIS climatology at SEDEE Boker are illustrated in Figures (12a, b).

401 MISR ‘matched’ AOD frequency show significant underestimation for AOD less than 0.2  
402 and an overestimation between 0.2 – 0.4 compared with AERONET retrievals. MISR



403 correctly captures the climatology for AOD events greater than 0.4. MISR 'matched' and 'all  
404 data' retrievals peaks at 0.2 producing high frequency of AOD oversampling compared to  
405 AERONET. MISR data retrievals do not capture the climatology for AOD less than 0.1 over  
406 this site coincident with what was previously observed over other sites. MODIS matched  
407 AERONET data underestimates frequency of AOD less than 0.2 while overestimates the  
408 frequencies between 0.2 - 0.6, and well match frequencies of higher AOD events larger than  
409 0.6. MODIS retrievals are characterized by two prominent peaks at 0.1 and 0.25 that are not  
410 found in the AERONET matched data.

411 At Sedee, MISR and MODIS retrievals are better in matching frequency of high AOD  
412 retrievals (greater than 0.4) than the frequency of low AOD. This could be an effect of  
413 possible long-range transport to Sedee Boker site (Farahat et al. 2016) along with complex  
414 mixtures of dust, pollution, smoke, and sea salt that could result in uncertainties in MISR and  
415 MODIS aerosol model selection.

416 In the summary, MISR tends to overestimate AOD  $> 0.4$  over Solar Village, Bahrain and  
417 underestimate AOD  $> 0.4$  over Cairo. MISR retrievals also match AOD  $> 0.4$  for Mezaria  
418 and Sedee Boker, while agree with AERONET over SAADA and Taman at all ranges of  
419 AOD. This could be expounded by insufficient particle absorption in MISR algorithm (Kahn  
420 et al., 2005). Spherical particle absorption is produced by externally mixing small black  
421 carbon particles.

422 Percentage of MISR, MODIS, and AERONET AOD greater than 0.4 recorded is shown in  
423 Table 4. Over Solar Village, both MISR and MODIS well capture high AOD greater than  
424 0.4 with very good agreement with the ground observations. Over Mezaria, both MISR and  
425 MODIS are over estimating the percentage of AOD greater than 0.4 by about 17.7 and 12.7  
426 percent respectively. MISR all data agrees well with AERONET all data in representing high  
427 AOD over Bahrain while MODIS shows significant under-representation of those events by

428 about 13 percent, less than reported by Bahrain AERONET station. At SAADA, MISR AOD  
429 agrees with AERONET in showing low percentage of AOD greater than 0.4, while MODIS  
430 retrievals overestimate percentage by about 24 percent. MISR AOD over Taman AERONET  
431 station shows very good agreement, while MODIS is slightly underestimating AOD. Among  
432 all seven sites considered in this study, Sedee Boker shows lowest occurrence of AOD greater  
433 than 0.4, which is confirmed by both MISR and MODIS retrievals. Cairo AERONET records  
434 the highest frequency of AOD > 0.4, however this is largely underestimated by both MISR  
435 and MODIS retrievals.

436 It can be concluded from the previous discussion that the atmosphere around SAADA,  
437 Taman, and Sedee Boker sites is relatively clean and aerosol loads are small compared to  
438 Solar Village, Mezaria, Bahrain, and Cairo, however this could be affected by the location  
439 where AERONET station is installed for example SAADA and Taman stations are installed  
440 in a remote mountainous region away from urbanization while Cairo station is installed in  
441 the middle of large residential region with significant local emissions.

442

#### 443 **Conclusion**

444 The performance of MODIS, MISR retrievals with corresponding AERONET  
445 measurements over different geographic locations in the Middle East and North Africa was  
446 investigated during 2000 – 2015.

447 Long-term observations show dissimilar AOD trends between MODIS/Aqua,  
448 MODIS/Terra, MISR and AERONET measurements. MODIS/Aqua matched AERONET  
449 retrievals show stable trend over all sites while, MODIS/Terra matched AERONET retrievals  
450 show significant downward trend indicating possible changes in the sensor performance.

451 MISR matched AERONET AOD data depict high correlation compared to  
452 AERONET indicating good agreement with ground observations with about 84 percent of  
453 AOD retrievals fall within the expected confidence range.

454 Consistency of MODIS and AERONET AOD vary based on the season, study area,  
455 and dominant aerosols type with about 64 percent of the retrieved AOD values fall within  
456 expected confidence range with the lowest performance over mixed particles regions.

457 Comparing satellites' AOD retrievals with corresponding AERONET measurements  
458 show that space-borne data retrievals accuracy can be affected by landscape, topology, and  
459 AOD range at which data is retrieved.

460 Few AERONET sites are verified where MISR and MODIS retrievals agree well with  
461 ground observations, while other sites only MISR or MODIS could correctly describe the  
462 climatology.

463 The AOD range at which MISR or MODIS could correctly describe ground  
464 observation is also investigated over different AERONET sites. Over Solar Village both  
465 MISR and MODIS tend to underestimate the frequency of low AOD and overestimate the  
466 frequency of high AOD compared to AERONET with MISR histograms show prominent  
467 peaks at 0.50 that matched AERONET data and 0.75 that could not be recorded in  
468 AERONET. MISR can capture the frequency of AOD greater than 1 mostly missed by  
469 MODIS. Both MISR and MODIS are found to provide good representation of the AOD  
470 climatology over the Solar Village site.

471 Similar to Solar Village, MISR underestimates frequency of lower AOD and  
472 overestimate frequencies of high AOD over Mezaria. MISR is able to correctly capture the  
473 frequency of AOD greater than 1, while MODIS retrievals are found to better represent the  
474 overall climatology. This is due to low number of MISR – matched AERONET retrievals

475 compared to MODIS over this site. Prominent peaks at 0.3 and 0.4 were observed in MODIS  
476 matched Mezaria retrievals compared to the climatology, which has a single peak at 0.30.

477         Large water body surrounding Bahrain makes MODIS data preferentially originate  
478 from the Arabian Gulf which produces an extremely large frequency of AOD at 0.1 not  
479 observed in AERONET measurements paired with an underestimation of AOD greater than  
480 0.3. Meanwhile, MISR retrievals agree well with AOD climatology over Bahrain.

481 MISR AOD retrievals slightly underestimate SAADA climatology while they show good  
482 agreement for AOD greater than 0.1. MODIS retrievals underestimate the frequency of AOD  
483 retrievals between 0.1 to 0.15, match climatology at AOD between 0.2 to 0.25, and  
484 overestimate the frequency of AOD greater than 0.3. SAADA site is characterized by small  
485 frequency of low AOD values and this could be related to the landscape nature surrounding  
486 Saada station. MISR is found to be outperforming over Saada and Taman stations which can  
487 be attributed to its viewing multispectral and multiangular capabilities over mountainous  
488 regions.

489         MISR retrievals well capture prominent peaks of AERONET data at 0.15 to 0.35  
490 with small underestimation observed at AOD greater than 0.3 over Cairo. Using either MISR  
491 matched data or MISR all data over Cairo was found to perform well in describing the  
492 climatology over this station. MODIS data retrievals are also able to well represent Cairo  
493 climatology with a high overestimation of AOD frequency between 0.05 to 0.2 and an  
494 underestimation of AOD larger than 0.4. While both MISR and MODIS well describe  
495 climatology over Cairo station, MODIS can correctly represent ground observations between  
496 0.2 to 0.4.

497 Over Sedee Boker both MISR and MODIS retrievals well describe the climatology however  
498 they are more successful in matching frequency of high AOD greater than 0.4.

499 Based on analysing frequency of AOD greater than 0.4, it was found that Saada, Taman, and  
500 Sedee Boker are having better air quality compared to other sites while Cairo was found to  
501 be the most polluted site.

502 Results presented in this study are important in providing a guideline for satellites retrievals  
503 end users on which sensor could provide reliable data over certain geographic location and  
504 AOD range.

505 Adjacent geographic location and local climate among sites does not always  
506 guarantee that same sensor will provide consistent retrievals over all sites. For example, Solar  
507 Village, and Bahrain AERONET are surrounded by large desert regions and sharing almost  
508 similar climatic conditions, but MODIS is found to be more successful in describing  
509 climatology over Solar Village than over Bahrain and this could be attributed to different  
510 factors related to surface reflection, cloud coverage, and the large water body surrounding  
511 Bahrain. Thus in order to decrease data uncertainty, it is important to determine which sensor  
512 provides best retrieval over certain geographic location and AOD range.

513

#### 514 **Acknowledgements**

515 The author would like to acknowledge the support provided by the Deanship of Scientific  
516 Research (DSR) at the King Fahd University of Petroleum and Minerals (KFUPM) for  
517 funding this work through project # IN161053. Portions of this work were performed at the  
518 Jet Propulsion Laboratory (JPL), California Institute of Technology, under a contract with  
519 the National Aeronautics and Space Administration. The author would like to thank Michael  
520 Garay (MJG) and Olga Kalashnikova (OVK) (JPL) for their suggestion of investigating  
521 satellites – AERONET matched data climatology, and discussion during the data analysis.  
522 The author would also like to thank Hesham El-Askary (Chapman University) for providing  
523 recommendation about AERONET data over North Africa and the Middle East as well as  
524 reviewing the English in the manuscript. We thank the MISR project for providing facilities,  
525 and supporting contributions of MJG and OVK. Finally, we thank the reviewers for  
526 suggestions, which improved the manuscript.

527

528

529 **Author Contributions:** Ashraf Farahat analysed the data, performed the statistical analysis  
530 and wrote the manuscript.

531

532 **Conflicts of Interest:** The authors declare no conflict of interest.

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794 **Tables' caption**

795 Table 1. Geographic location of the AERONET sites used in this study

796 Table 2. Statistics for dust sites, R: correlation coefficient, RMSE: Root Mean Square

797 deviation; G-fraction: good fraction; N: number of observations

798 Table 3. Statistics for biomass and mixed sites, parameters as in Table 3. Caption.

799 Table 4. MISR coverage for six days of major dust activity over the Arabian Peninsula

800 during March 2009.

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819 **Figures caption**

820 Figure 1. Location of the AERONET stations over North Africa and the Middle East. The  
821 numbers on the map indicates the site location as 1: Saada, 2: Tamanrasset\_INM, 3: Cairo,  
822 4: Sede Boker, 5: Solar Village, 6: Mezaira, 7: Bahrain.

823 Figure 2. Scatter plot of MISR AOD versus AERONET AOD based on seasons and  
824 aerosols categorization.

825 Figure 3. Scatter plot of MODIS AOD versus AERONET AOD based on seasons and  
826 aerosols categorization.

827 Figure 4. Time series of monthly mean AOD derived from MODIS/Aqua, MODIS/Terra,  
828 MISR and AERONET over a) dust b) biomass and c) mixed dominated aerosol regions.

829 Figure 5. Long-term AOD difference for MODIS/Terra and MODIS/Aqua over the Solar  
830 Village and Taman sites.

831 Figure 6. Histogram of the MISR, MODIS and Solar Village AERONET measurements a)  
832 MISR b) MODIS data retrievals.

833 Figure 7. Histogram of the MISR, MODIS and Mezaria AERONET measurements a)  
834 MISR b) MODIS data retrievals.

835 Figure 8. Histogram of the MISR, MODIS and Bahrain AERONET measurements a) MISR  
836 b) MODIS data retrievals.

837 Figure 9. Histogram of the MISR, MODIS and SAADA AERONET measurements a)  
838 MISR b) MODIS data retrievals.

839 Figure 10. Histogram of the MISR, MODIS and Taman AERONET measurements a)  
840 MISR b) MODIS data retrievals.

841 Figure 11. Histogram of the MISR, MODIS and SEDEE Boker AERONET measurements  
842 a) MISR b) MODIS data retrievals.

843 Figure 12. Histogram of the MISR, MODIS and Cairo AERONET measurements a) MISR

844 b) MODIS data retrievals.

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Table 1.

<b>Location name</b>	<b>Lon./Lat.</b>	<b>Measurement period</b>
<b>Solar Village</b>	24.907° N/46.397° E	2000-2015
<b>Mezaria</b>	23.105° N/53.755° E	2004-2015
<b>Bahrain</b>	26.208° N/50.609° E	2000-2006
<b>Saada</b>	31.626° N/8.156° W	2003-2015
<b>Taman</b>	22.790° N/5.530° E	2000-2015
<b>Cairo (EMA_2)</b>	30.081° N/31.290° E	2010 -2017
<b>Sede Boker</b>	30.855° N/34.782 ° E	2000-2015

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<b>AERONET</b>	<b>Sensor</b>	<b>Season</b>	<b>Mean Value</b>	<b>N</b>	<b>R</b>	<b>Gfraction (%)</b>	
<b>Site</b>			<b>AERONET</b>	<b>Satellite</b>			
<b>Solar Village</b>	<b>MISR</b>	<b>DJF</b>	0.18±0.15	0.23±0.13	24	0.94	79.1
		<b>MAM</b>	0.45±0.21	0.47±0.20	43	0.94	86.0
		<b>JJA</b>	0.39±0.16	0.42±0.16	57	0.90	82.4
		<b>SON</b>	0.25±0.14	0.29±0.12	50	0.99	82.0
	<b>MODIS</b> <b>Terra</b>	<b>DJF</b>	0.27±0.19	0.33±0.17	1500	0.48	51.80
		<b>MAM</b>	0.36±0.24	0.26±0.17	389	0.68	90.23
		<b>JJA</b>	0.34±0.17	0.42±0.19	429	0.41	54.31
		<b>SON</b>	0.22±0.10	0.36±0.12	471	0.51	28.87
<b>Mezaria</b>	<b>MISR</b>	<b>DJF</b>	0.17±0.09	0.23±0.07	53	0.89	50.9
		<b>MAM</b>	0.34±0.18	0.37±0.18	41	0.90	78.0
		<b>JJA</b>	0.49±0.20	0.47±0.21	51	0.85	92.1
		<b>SON</b>	0.26±0.09	0.30±0.12	53	0.87	88.2
	<b>MODIS</b> <b>Terra</b>	<b>DJF</b>	0.32±0.15	0.35±0.19	198	0.86	74.74
		<b>MAM</b>	0.44±0.33	0.45±0.27	115	0.92	78.07
		<b>JJA</b>	0.39±0.14	0.43±0.20	89	0.81	71.91
		<b>SON</b>	0.28±0.13	0.30±0.16	97	0.87	77.31
<b>Bahrain</b>	<b>MISR</b>	<b>DJF</b>	0.19±0.10	0.30±0.10	9	0.73	33.3
		<b>MAM</b>	0.47±0.20	0.67±0.05	7	0.89	28.5
		<b>JJA</b>	0.45±0.21	0.74±0.21	21	0.69	23.8
		<b>SON</b>	0.32±0.13	0.45±0.16	22	0.98	45.4
	<b>MODIS</b> <b>Terra</b>	<b>DJF</b>	0.42±0.29	0.20±0.19	121	0.41	93.38
		<b>MAM</b>	0.50±0.28	0.13±0.15	25	0.26	96.00
		<b>JJA</b>	0.55±0.26	0.31±0.27	42	0.50	88.09
		<b>SON</b>	0.35±0.14	0.21±0.12	29	0.32	93.10



Table 3.

AERONET Site	Method	Season	Mean Value	N	R	Gfraction (%)		
			AERONET	Satellite				
SAADA	MISR	DJF	0.07±0.02	0.07±0.02	43	0.93	100.0	
		MAM	0.17±0.10	0.17±0.09	47	0.89	93.6	
		JJA	0.30±0.14	0.31±0.14	53	0.93	93.1	
		SON	0.14±0.07	0.13±0.06	51	0.94	96.0	
	MODIS	DJF	0.23±0.16	0.32±0.21	550	0.57	57.8	
		MAM	0.24±0.18	0.39±0.23	90	0.43	44.4	
		JJA	0.30±0.17	0.45±0.18	201	0.40	45.2	
		SON	0.19±0.13	0.22±0.14	162	0.71	72.3	
	Taman	MISR	DJF	0.07±0.10	0.09±0.06	69	0.92	85.5
			MAM	0.22±0.18	0.25±0.22	86	0.97	81.3
			JJA	0.42±0.31	0.45±0.28	57	0.85	78.9
			SON	0.14±0.11	0.15±0.10	72	0.94	95.8
MODIS		DJF	0.19±0.22	0.18±0.16	319	0.67	81.8	
		MAM	0.24±0.19	0.22±0.17	67	0.55	83.5	
		JJA	0.37±0.32	0.29±0.20	69	0.69	84.0	
		SON	0.14±0.14	0.13±0.10	117	0.54	84.6	
Cairo	MISR	DJF	0.33±0.17	0.17±0.09	15	0.94	100.0	
		MAM	0.35±0.13	0.33±0.15	39	0.99	82.0	
		JJA	0.35±0.09	0.27±0.08	61	0.99	96.7	
		SON	0.37±0.14	0.28±0.13	23	0.97	78.2	
	MODIS	DJF	0.33±0.16	0.20±0.11	158	0.30	95.5	
		MAM	0.32±0.16	0.12±0.08	39	0.25	100.0	
		JJA	0.35±0.14	0.28±0.07	58	0.17	94.8	
		SON	0.38±0.19	0.20±0.09	29	0.07	93.8	

	<b>DJF</b>	0.11±0.06	0.13±0.05	10	0.87	90.0
	<b>MAM</b>	0.21±0.13	0.24±0.13	76	0.68	75.0
<b>MISR</b>	<b>JJA</b>	0.16±0.08	0.21±0.08	142	0.85	66.9
	<b>SON</b>	0.162±0.07	0.20±0.06	54	0.89	79.6
<b>SEDEE_BOKER</b>	<b>DJF</b>	0.16±0.12	0.23±0.14	1312	0.36	53.5
	<b>MAM</b>	0.21±0.18	0.24±0.19	338	0.34	65.6
<b>MODIS</b>	<b>JJA</b>	0.16±0.09	0.33±0.13	392	0.27	17.3
<b>Terra</b>	<b>SON</b>	0.16±0.09	0.23±0.12	477	0.46	58.4

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Table 4.

	AERONET		MISR		MODIS	
	N	AOD	N	AOD	N	AOD
		% > 0.4		% > 0.4		% > 0.4
<b>Solar</b>	3893	27.17	684	32.8	2789	30.1
<b>Village</b>						
<b>Mezaria</b>	2245	28.01	547	45.7	498	40.7
<b>Bahrain</b>	1116	31.36	676	35.8	217	18.4
<b>SAADA</b>	2974	10.32	667	11.5	1004	34.6
<b>Taman</b>	798	15.78	845	22.6	572	9.4
<b>Cairo</b>	2222	38.79	620	17.7	284	4.2
<b>SEDEE</b>	5722	4.28	675	9.0	2519	12.8

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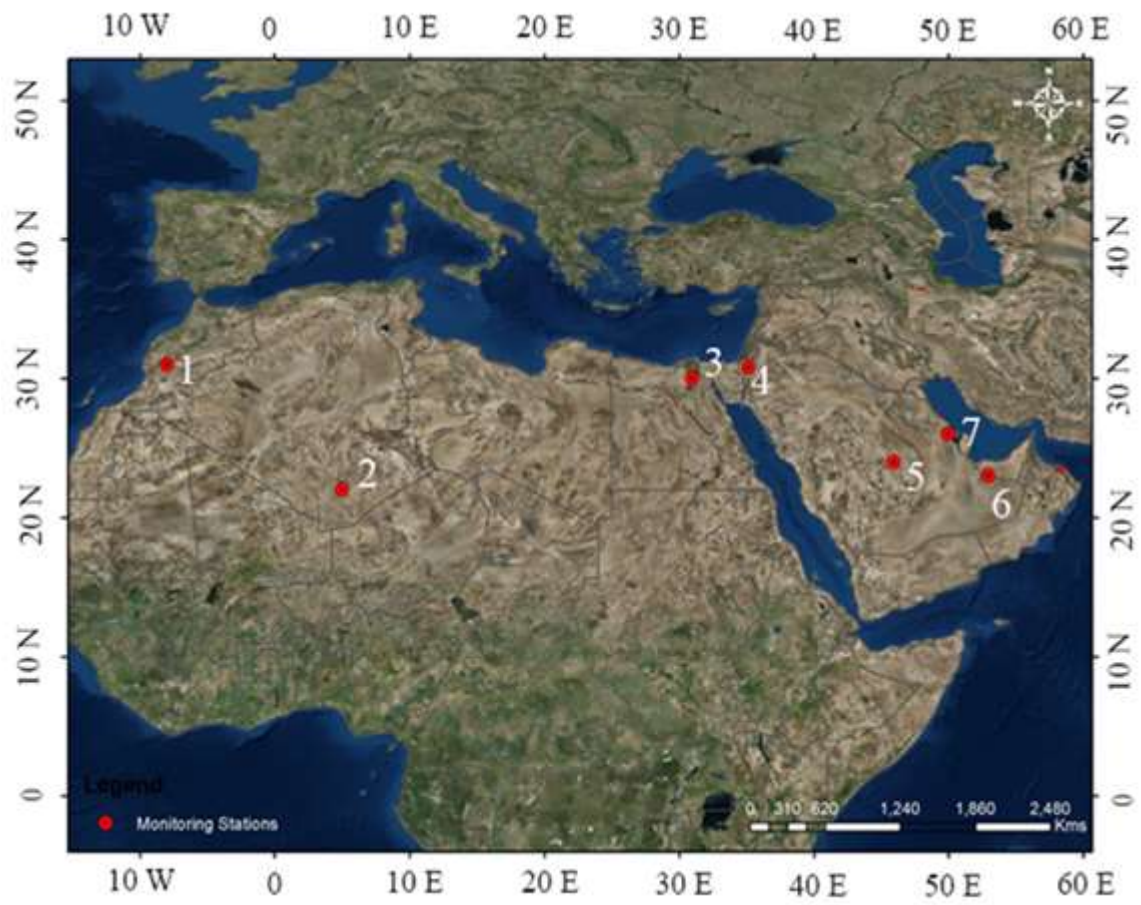
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928 Figure 1.

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Region 1

Region 2

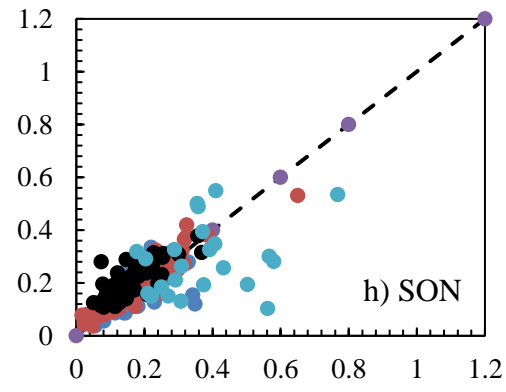
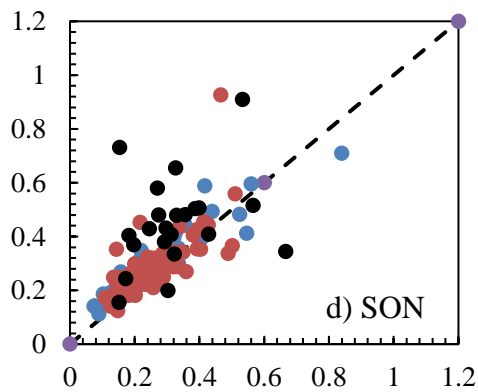
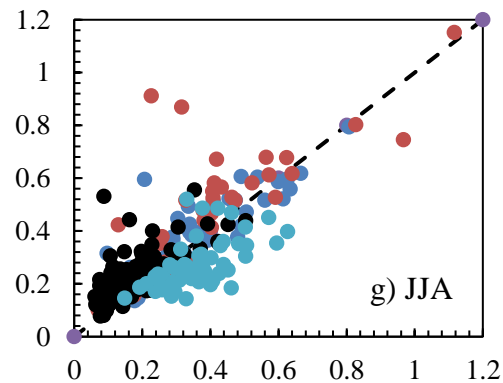
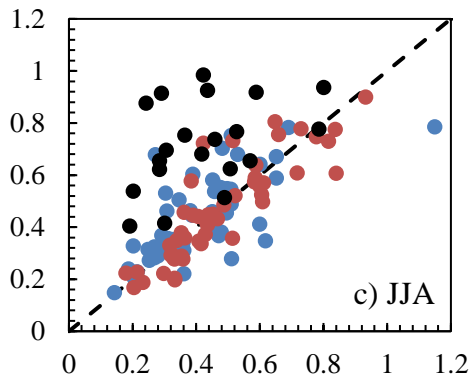
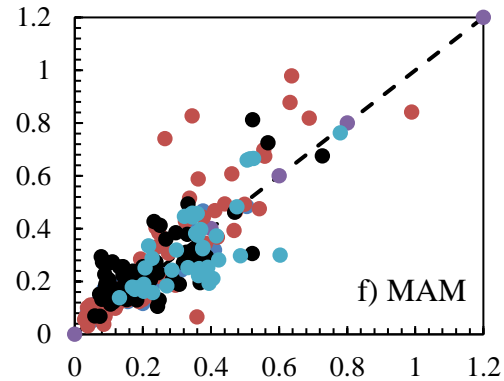
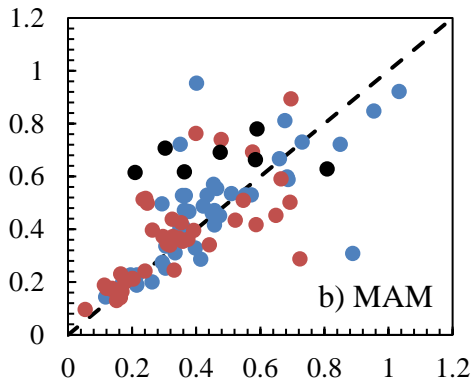
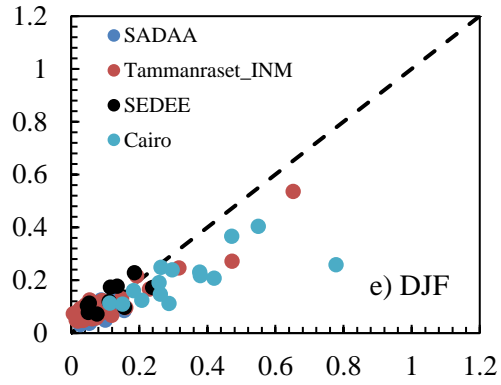
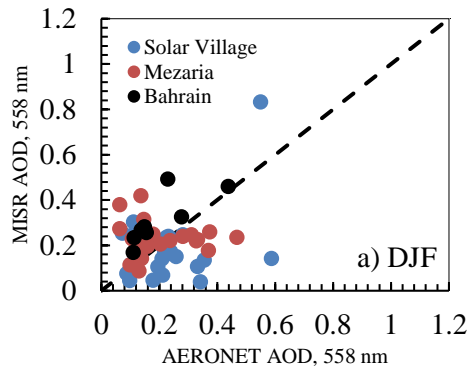


Figure 2

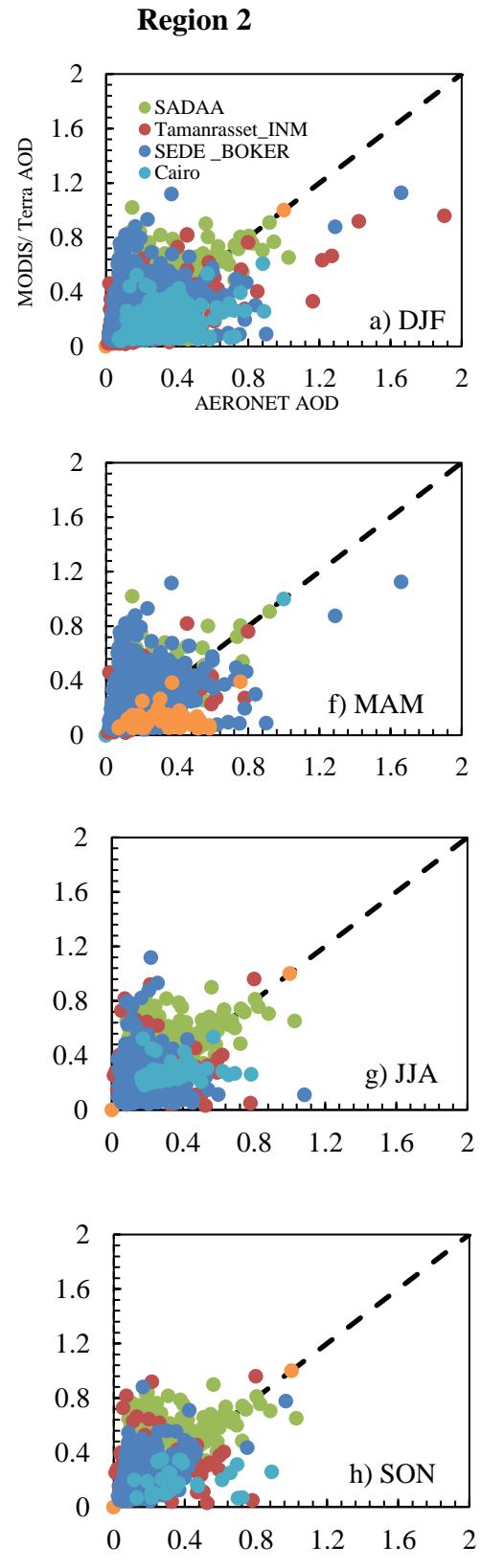
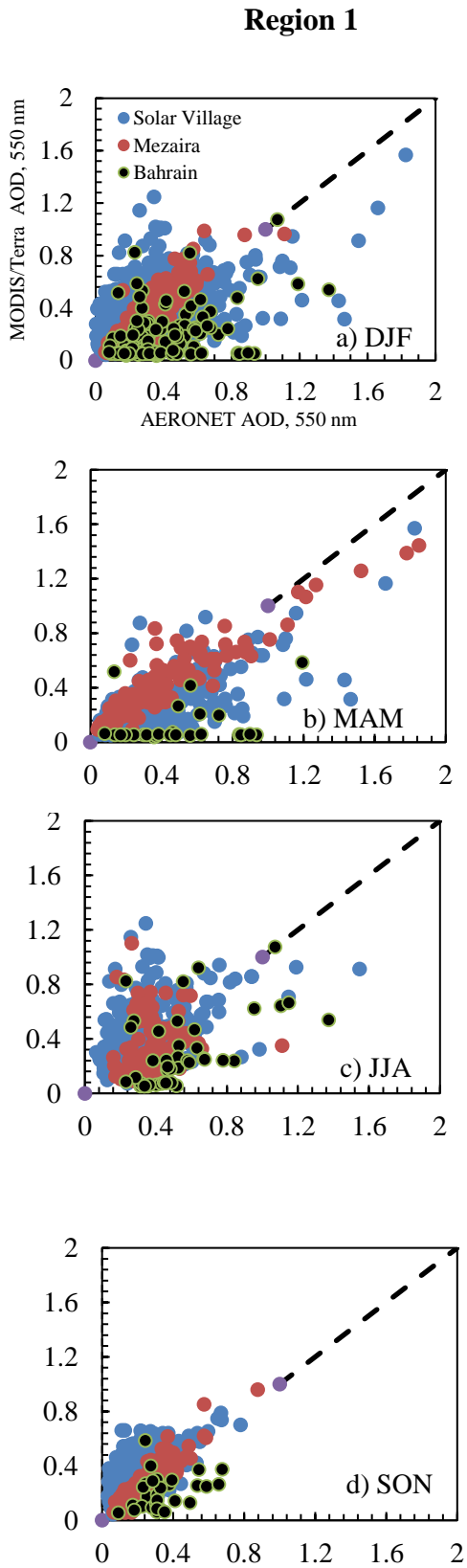
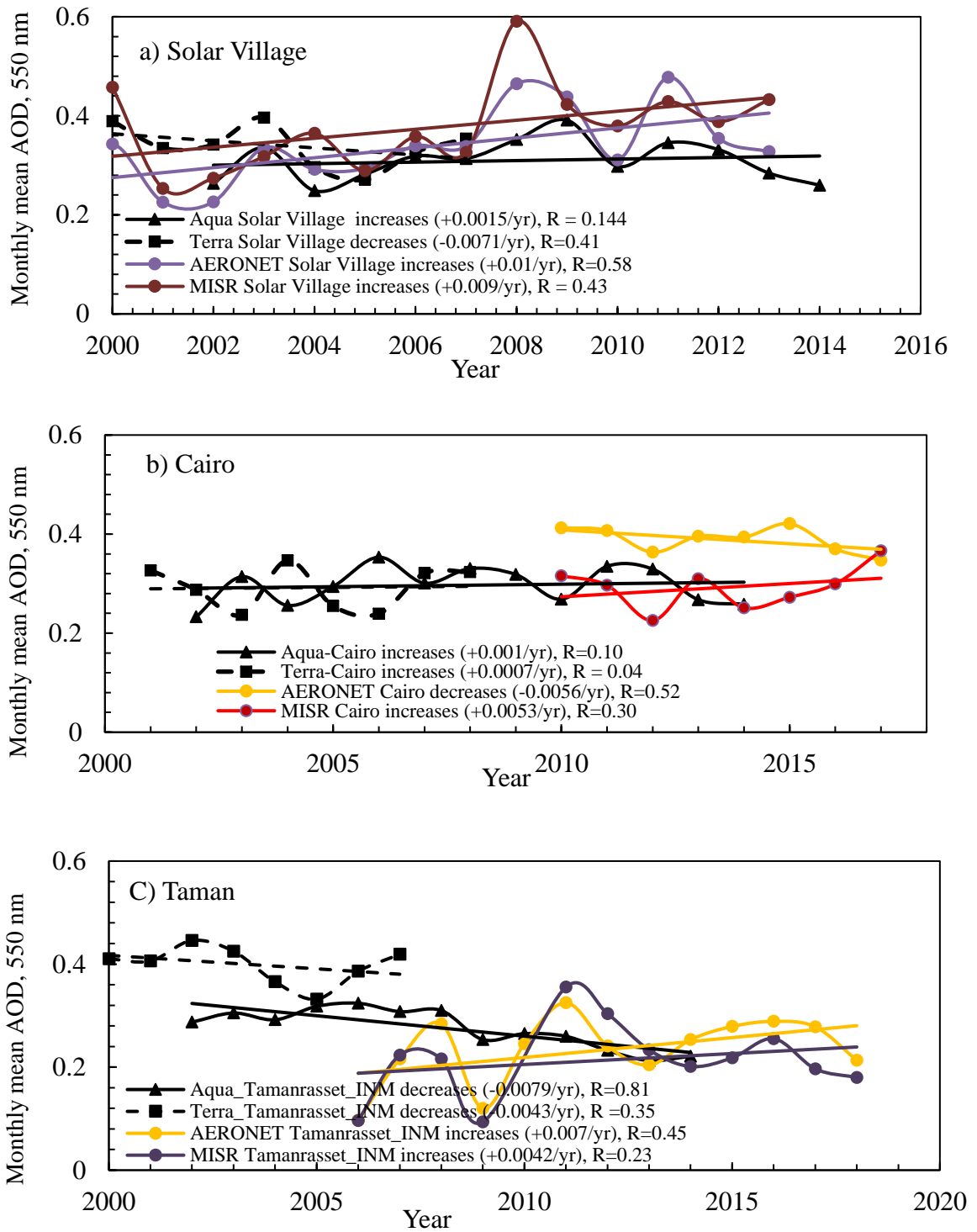


Figure 3.



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Figure 4.

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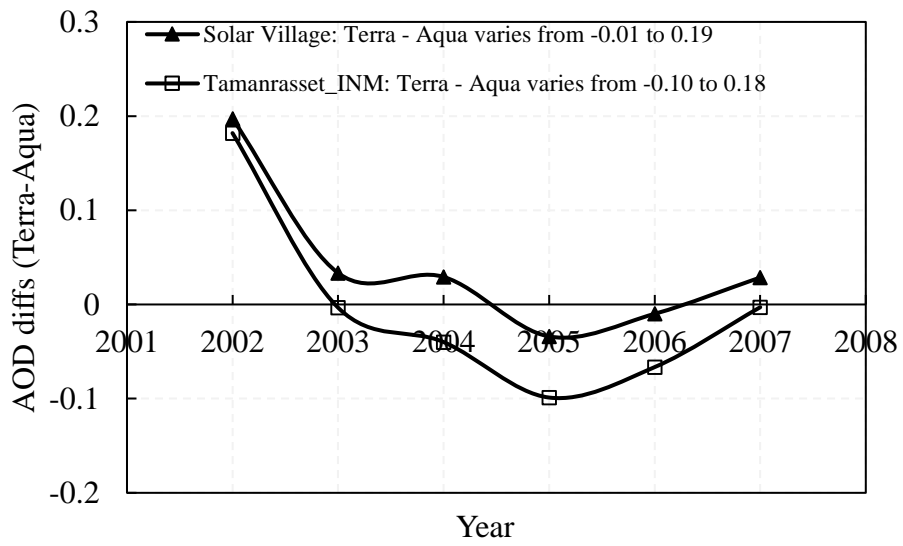
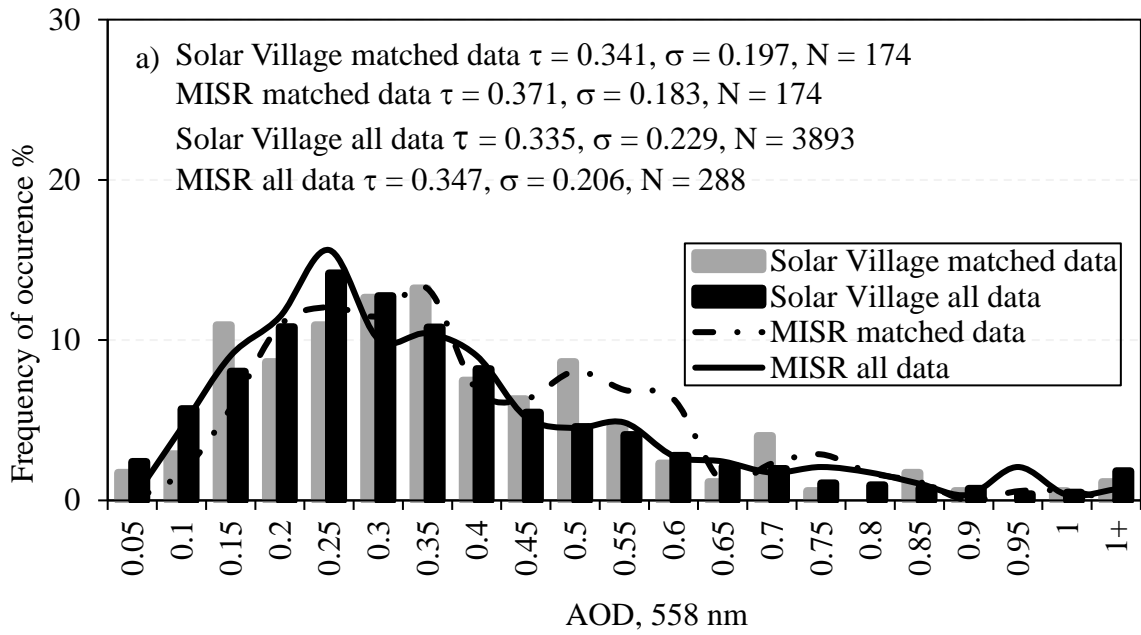


Figure 5.

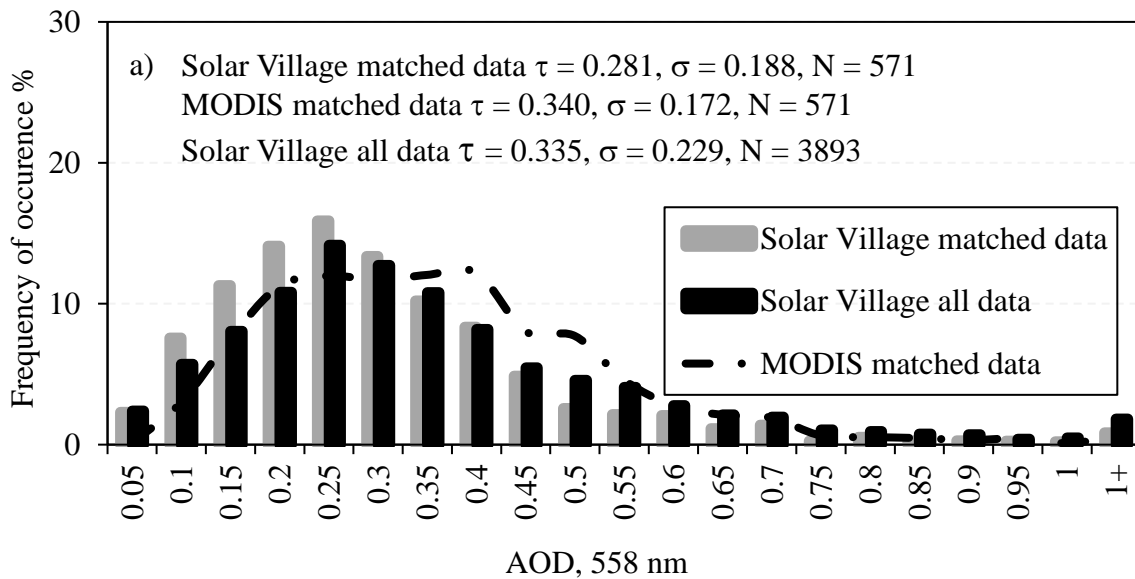
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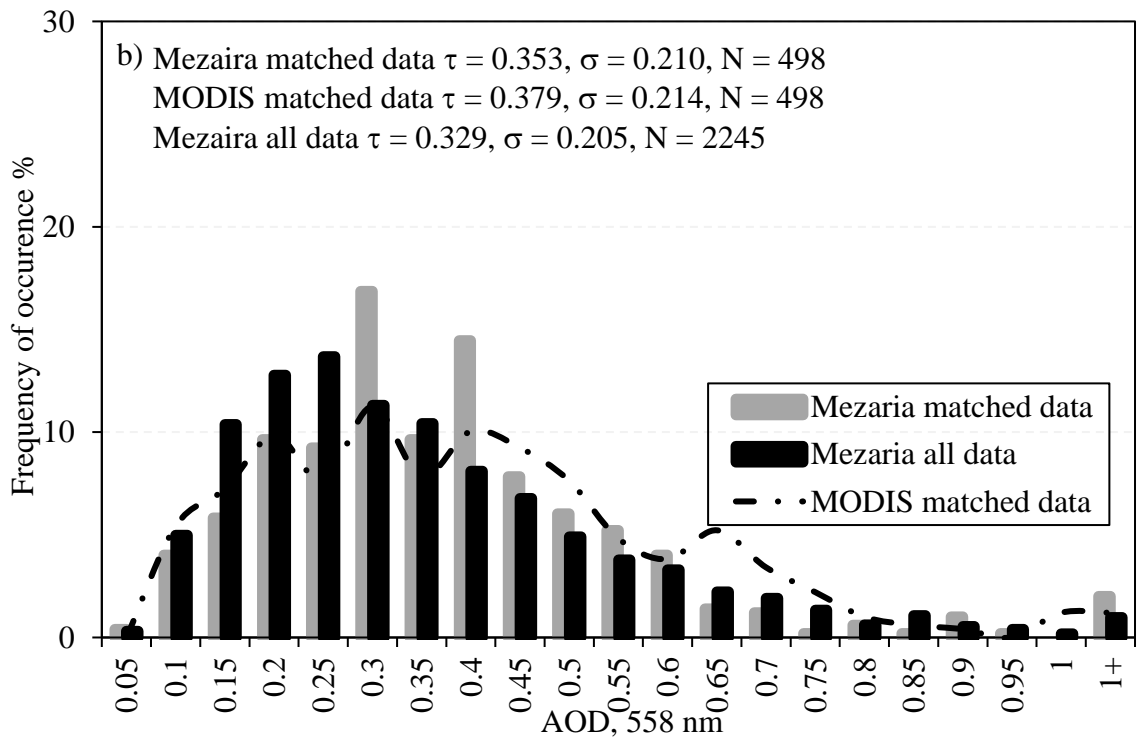
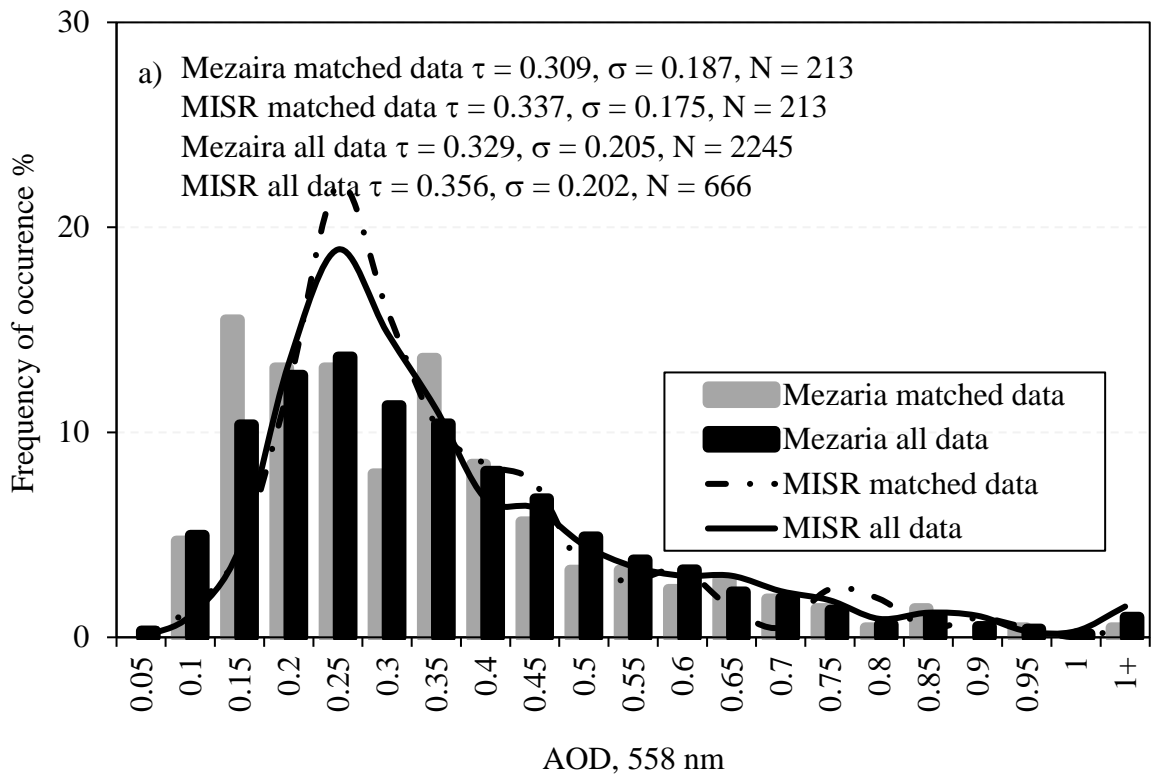


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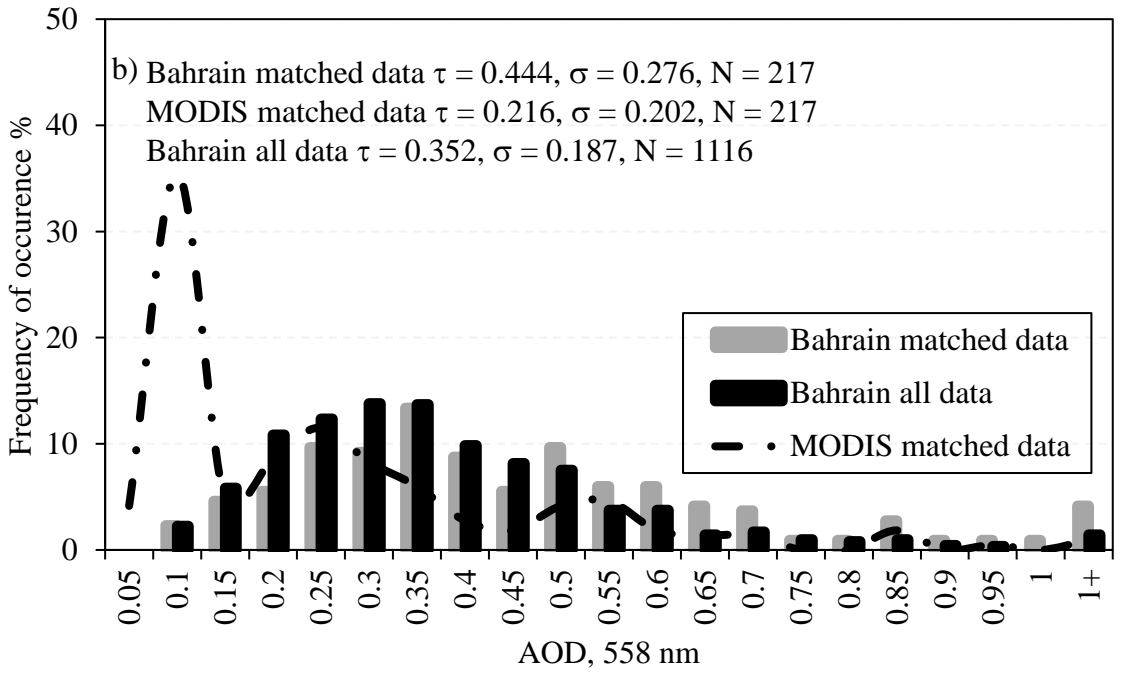
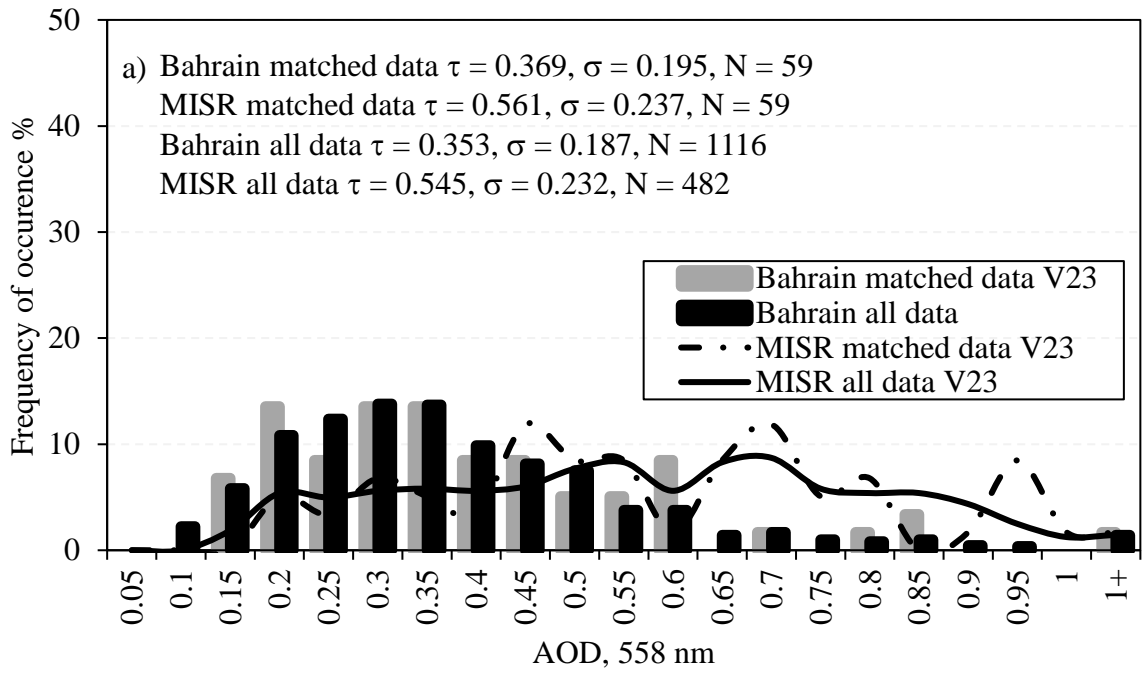


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Figure 6.



982 Figure 7.

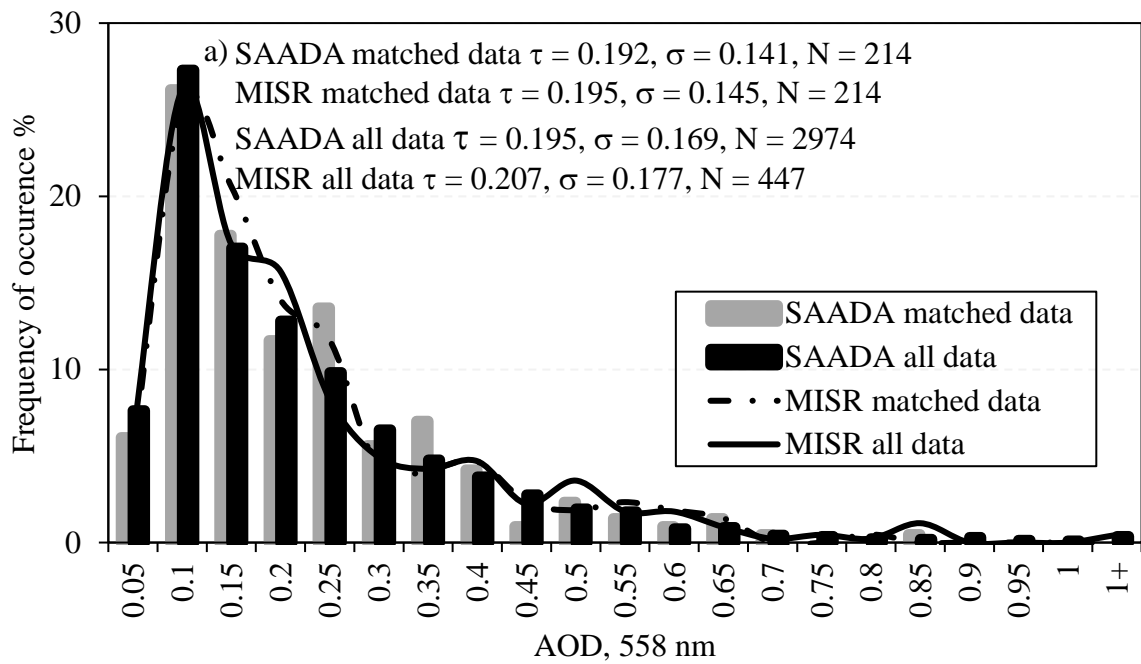


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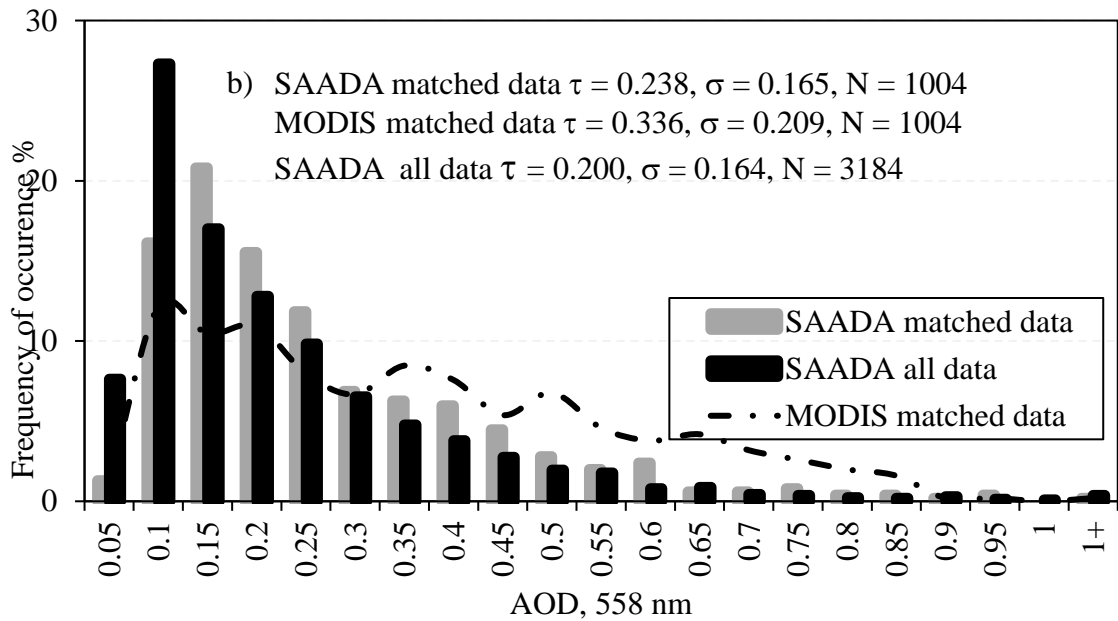
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985 Figure 8.

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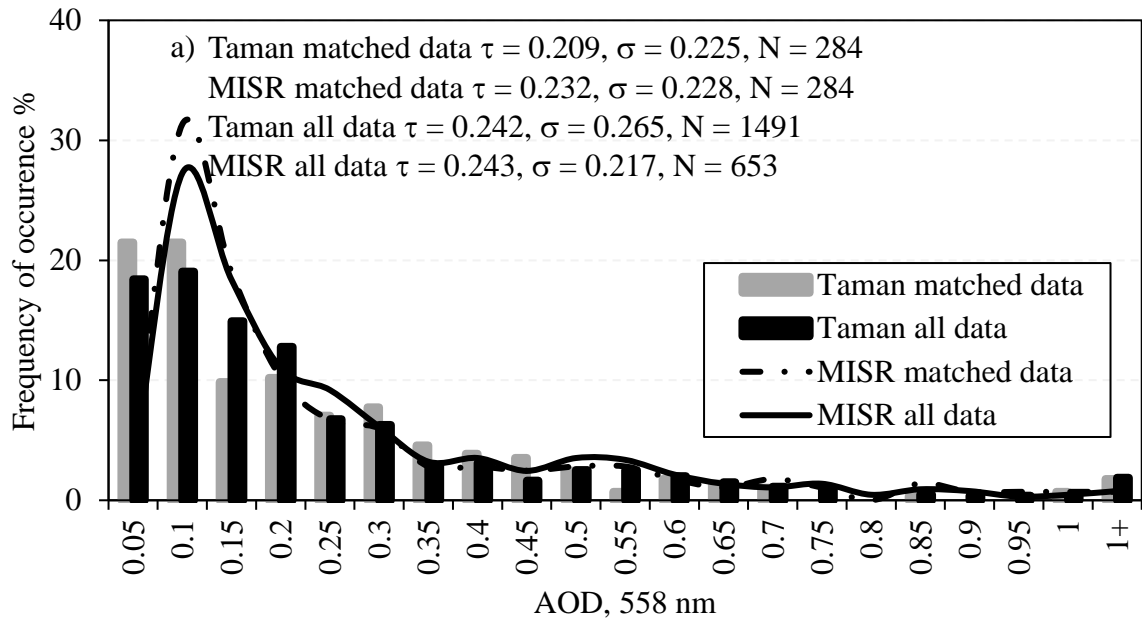
989 Figure 9.

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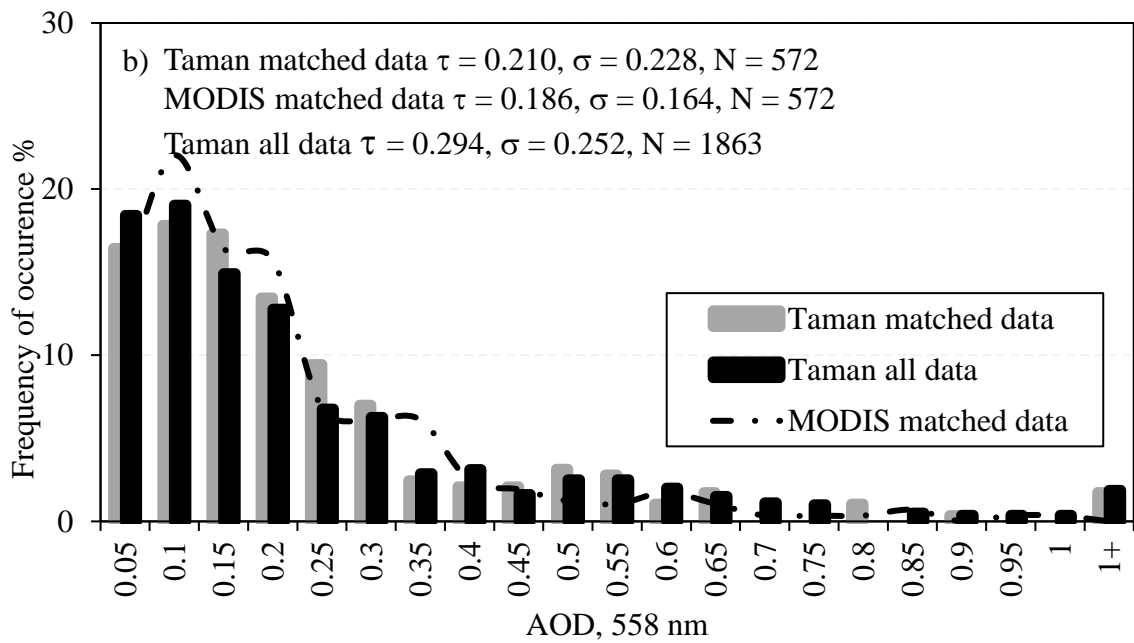
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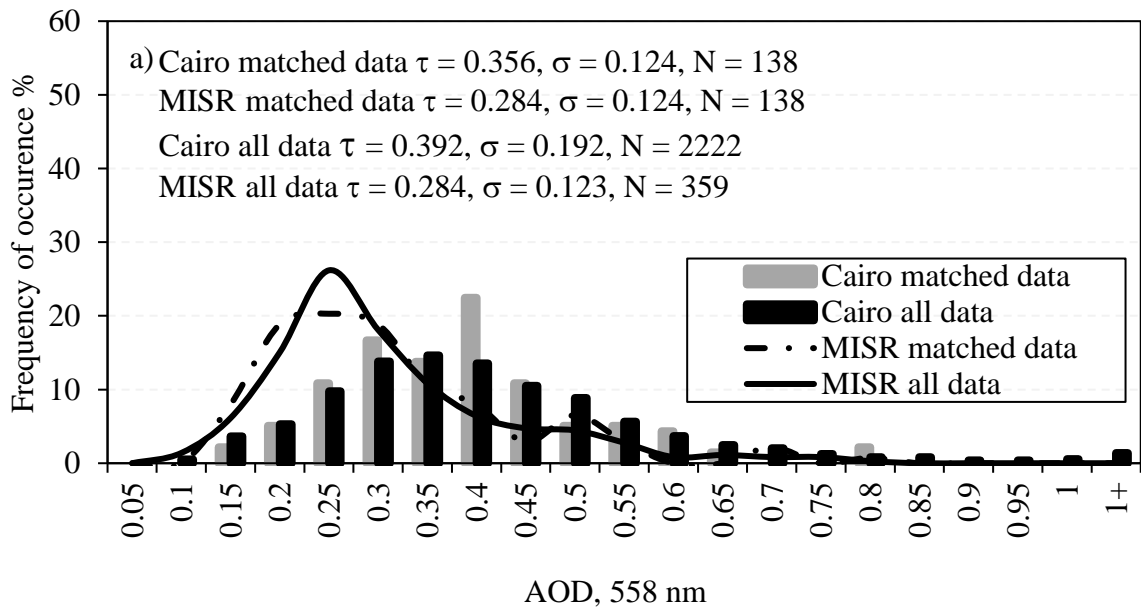
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996 Figure 10.

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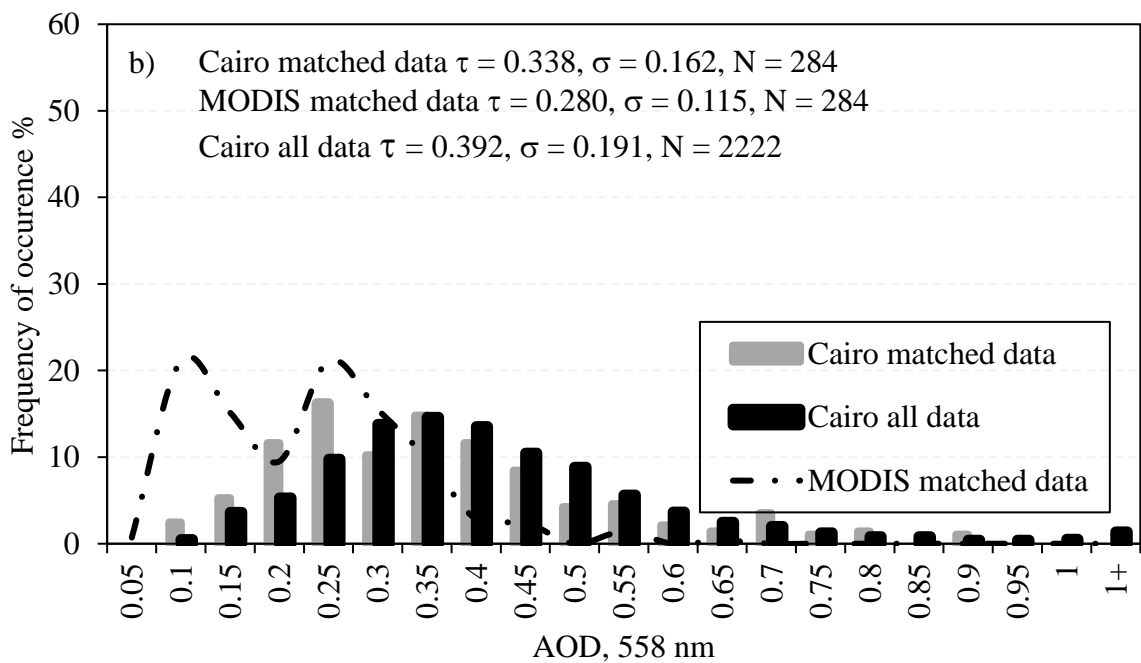
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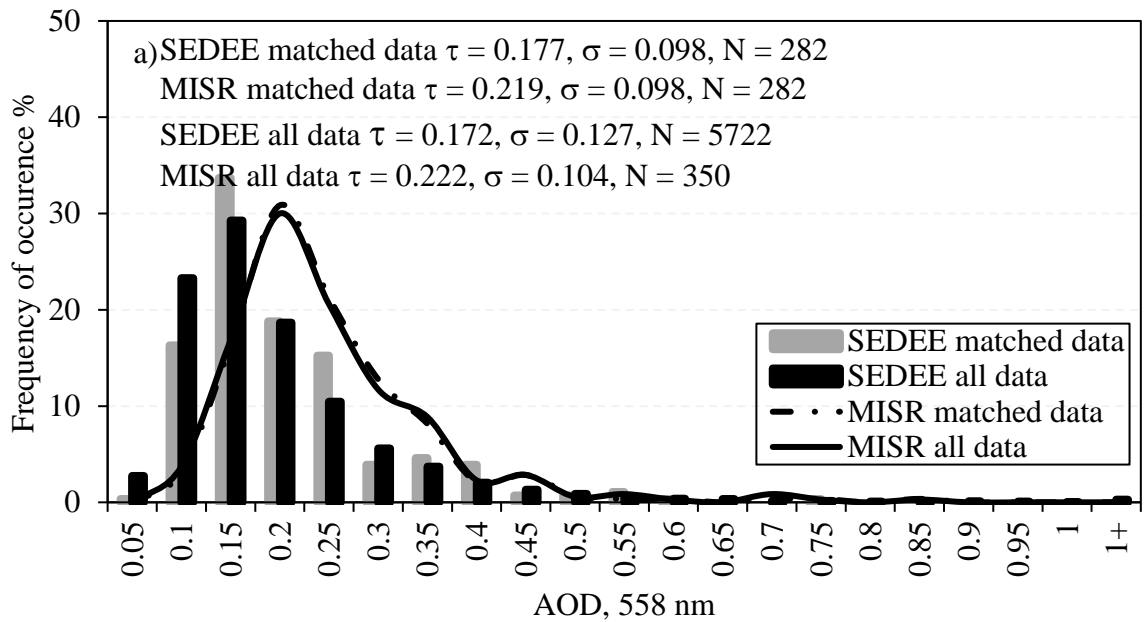


1002 Figure 11.

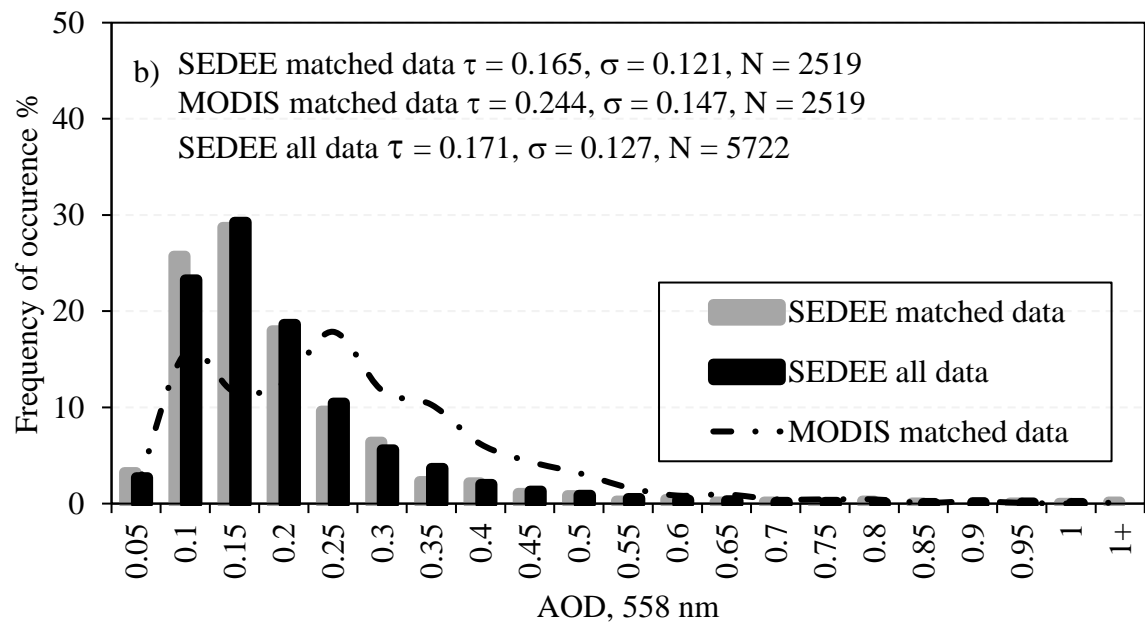
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1008 Figure 12.