



1     **Intra-annual variations of regional aerosol optical depth, vertical distribution, and**  
2     **particle types from multiple satellite and ground-based observational datasets**

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10



## 11 Abstract

12 The relatively short lifetimes of aerosols in the atmosphere result in climatic and health  
13 effects that are strongly dependent on intra-annual variations in particle concentrations. While  
14 many studies have examined the seasonal and diurnal variations of regional aerosol optical depth  
15 (AOD), understanding the temporal variations in aerosol vertical distribution and particle types is  
16 also important for accurate computation of aerosol radiative effects. In this paper, we combine  
17 the observations from four satellite-borne sensors and ground-based AOD and fine particle  
18 (PM<sub>2.5</sub>) measurements to investigate the seasonal and diurnal variations of aerosol column  
19 loading, vertical distribution, and particle types over three populous regions: the Eastern United  
20 States (EUS), Western Europe (WEU), and Eastern and Central China (ECC). In all three regions,  
21 column AOD, as well as AOD higher than 800 m above ground level, peaks in summer/spring  
22 probably due to accelerated formation of secondary aerosols and hygroscopic growth. However,  
23 AOD at height below 800 m mostly peaks in winter except that a second maximum in summer  
24 occurs over the EUS region, which is consistent with observed temporal trends in surface PM<sub>2.5</sub>  
25 concentrations. AOD due to fine particles (< 0.7 μm diameter) is much larger in spring/summer  
26 than in winter over all three regions, whereas coarse mode AOD (> 1.4 μm diameter) generally  
27 shows less variability, except for the ECC region where a peak occurs in spring, consistent with  
28 the prevalence of airborne dust during this season. When aerosols are classified according to  
29 sources, the dominant type is associated with anthropogenic air pollution, which has a similar  
30 seasonal pattern as total AOD. Dust and sea-spray aerosols in the WEU region peak in summer  
31 and winter, respectively, but do not show an obvious seasonal pattern in the EUS region. Smoke  
32 aerosols, as well as absorbing aerosols, present an obvious unimodal distribution with a  
33 maximum occurring in summer over the EUS and WEU regions, whereas they follow a bimodal



34 distribution with peaks in August and March (due to crop residue burning) over the ECC region.  
35 In general, the nighttime-daytime AOD difference is more positive in summer than in winter,  
36 likely attributable to a larger diurnal temperature range in summer. Smoke AOD is much higher  
37 in the nighttime than in the daytime. The results of this study can help to improve the current  
38 estimates of the climatic and health impacts of aerosols.

39

## 40 **1 Introduction**

41 Aerosols have adverse effects on human health (Lelieveld et al., 2015) and play a key  
42 role in Earth's climate through aerosol-radiation interactions (McCormick and Ludwig, 1967)  
43 and aerosol-cloud interactions (Twomey, 1977; Albrecht, 1989; Garrett and Zhao, 2006).  
44 Compared with long-lived climate forcers such as CO<sub>2</sub>, aerosols have relatively short lifetimes  
45 and hence large spatiotemporal variability (Unger et al., 2008; Shindell et al., 2009). While the  
46 climatic effects of CO<sub>2</sub> are mainly induced by inter-annual concentration changes, the climatic  
47 and health effects of aerosols also strongly depend on their intra-annual (seasonal and diurnal)  
48 variability.

49 Aerosol optical depth (AOD) has been widely used to represent the column aerosol  
50 loading and to assess the aerosol impacts on radiation, clouds, and precipitation (Ma et al., 2014;  
51 Niu and Li, 2012; Zhao et al., 2018; Song et al., 2017). However, the wide ranges of particle  
52 scattering and absorption properties mean that even for the same AOD, different aerosol  
53 components have different effects on not only the magnitude, but also the sign, of aerosol  
54 radiative forcing (IPCC, 2013; Gu et al., 2006). IPCC (2013) estimates that the historical global  
55 mean direct radiative forcings due to sulfate, organic carbon (OC), black carbon (BC), and  
56 mineral dust are  $-0.40$ ,  $-0.19$ ,  $+0.36$ , and  $-0.10$  W m<sup>-2</sup>, respectively. Furthermore, absorbing



57 and non-absorbing aerosols have been found to have very different impacts on the development  
58 of convective clouds (Massie et al., 2016; Ramanathan et al., 2005; Rosenfeld et al., 2008).  
59 Besides aerosol type, perturbation of aerosol vertical distribution influences the vertical profile  
60 of heating rate (Johnson et al., 2008; Guan et al., 2010; Zhang et al., 2013), which subsequently  
61 modifies the atmospheric stability and convective strength (Ramanathan et al., 2007), with  
62 potential changes in regional circulation (Ramanathan et al., 2001) and cloud cover (Johnson et  
63 al., 2004). Understanding aerosol variability as a function of height is also important because the  
64 health impacts of aerosols are only associated with those present near the surface, where they are  
65 inhaled. For these reasons, systematic analyses of the intra-annual variations of aerosol vertical  
66 distribution and particle types, in addition to total column AOD, are necessary to improve our  
67 understanding of aerosol climatic and health effects.

68 Numerous studies have investigated the seasonal variations of AOD at global and  
69 regional scales using satellite observations (e.g., Kim et al., 2007; Song et al., 2009; Mehta et al.,  
70 2016; Mao et al., 2014). By comparison, most previous studies of the temporal variations of  
71 aerosol vertical distributions and aerosol types have been confined to only a few sites due to  
72 coverage limitations associated with reliance on ground-based instruments (e.g., Liu et al., 2012;  
73 Matthias et al., 2004). Despite continuous advancement of remote sensing technology and  
74 emergence of new spaceborne sensors, only limited number of studies have utilized satellite  
75 observations to examine the seasonal and/or diurnal variations of aerosol profiles and/or types at  
76 regional or larger scales (Huang et al., 2013; Kahn and Gaitley, 2015; Yu et al., 2010; Li et al.,  
77 2016). Huang et al. (2013) analyzed the seasonal and diurnal variations of aerosol extinction  
78 profile and type distribution using 5-year observations from the Cloud-Aerosol Lidar and  
79 Infrared Pathfinder Satellite Observations (CALIPSO). Kahn and Gaitley (2015) examined the



80 spatiotemporal variations of aerosol types retrieved by the Multi-angle Imaging  
81 SpectroRadiometer (MISR). Different satellite-borne sensors, such as MISR, CALIPSO, and  
82 Moderate resolution Imaging Spectroradiometer (MODIS), employ different principles of  
83 measurement and retrieval, and therefore provide different sensitivities to column AOD, aerosol  
84 types, and vertical profiles. Therefore, integration of data from multiple satellites and ground-  
85 based observational networks makes it possible to deepen our understanding of the intra-annual  
86 variations of aerosol loadings, profiles, and types.

87 In this study, we investigate the seasonal and diurnal variations of aerosol column loading,  
88 vertical distribution, and particle types using multiple satellite and ground-based observational  
89 datasets covering the period from 2007 to 2016. The purpose is to assess the consistency among  
90 various datasets and provide a comprehensive characterization of aerosol properties in polluted  
91 regions to facilitate future studies of aerosol climate effects and local air quality issues. The data  
92 are from MISR, MODIS, CALIPSO, Aerosol Robotic Network (AERONET), and surface PM<sub>2.5</sub>  
93 monitors. Consistent with our previous study (Zhao et al., 2017), we selected three populous  
94 regions which have experienced substantial anthropogenic pollution (Wang et al., 2017; Wang et  
95 al., 2014) and have received considerable attention in other climate studies: the Eastern United  
96 States (EUS; 29°-45° N, 70°-98° W), Western Europe (WEU; 37°-59° N, 10° W-17° E), and  
97 Eastern and Central China (ECC; 21°-41° N, 102°-122° E). The geographical boundaries of these  
98 regions are shown in Fig. 1.

99 **2 Data and Methods**

## 100 2.1 Satellite data

101 We obtain retrievals of total column AOD as well as AOD for various height ranges and  
102 aerosol types from MISR (flying on the Terra satellite), MODIS (Terra and Aqua), and the  
103 Cloud-Aerosol Lidar with Orthogonal Polarization (CALIOP) on CALIPSO.

104 MISR observes the Earth with moderately high spatial resolution (275 m to 1.1 km) at 9  
105 along-track viewing angles in each of 4 visible/near-infrared spectral bands, which enables the  
106 partitioning of AOD by particle type over both land and ocean, in addition to retrieval of total  
107 AOD (Kahn and Gaitley, 2015; Kahn et al., 2001). Its observations provide near-global coverage  
108 every 9 days (Diner et al., 1998). We make use of the Level 3 monthly global aerosol product  
109 (MIL3MAE) version F15\_0031, which is generated at a spatial resolution of  $0.5^\circ \times 0.5^\circ$ . The  
110 variables used in the analysis are total AOD at 555 nm as well as AODs for six aerosol  
111 components, namely small ( $< 0.7 \mu\text{m}$  diameter), medium (0.7-1.4  $\mu\text{m}$  diameter), large ( $> 1.4 \mu\text{m}$   
112 diameter), spherical, non-spherical, and absorbing. Based on comparison with ground-based  
113 AERONET measurements, the errors in MISR AOD data are on the order of  $\pm 0.05$  or  $\pm(0.20 \times$   
114 AOD), whichever is larger (Kahn et al., 2005; Kahn et al., 2010). In addition, retrieval of MISR  
115 particle property information from individual retrievals is considered to be reliable when AOD  $>$   
116 0.15, and has diminished sensitivity at smaller AOD (Kahn and Gaitley, 2015; Kahn et al., 2010).  
117 In this study we use only monthly mean values, for which the uncertainties are expected to be  
118 smaller than those for individual retrievals. Note that we did not do a relative humidity (RH)  
119 correction to AOD retrievals from MISR as well as other sensors. The seasonal and diurnal  
120 variations of AOD represent an integrated effect of variations in aerosol abundance, vertical  
121 distribution, chemical constituents, and meteorological conditions.



122 The MODIS sensors onboard the Terra and Aqua satellites observe the Earth with  
123 multiple wavelength bands over a 2330 km swath (King et al., 2003), which provides near-daily  
124 global coverage. In this study we obtain column AOD data at 550 nm with a  $1^\circ \times 1^\circ$  resolution  
125 from the Level 3 monthly atmosphere products Collection 6 (MOD08 and MYD08 for the Terra  
126 and Aqua platforms, respectively). Comparison studies with AERONET have estimated the  
127 accuracy of AOD retrievals to be about  $\pm(0.05 + 0.15 \times \text{AOD})$  over land and  $\pm(0.03 + 0.05 \times$   
128  $\text{AOD})$  over ocean (Levy et al., 2010; Remer et al., 2005). For both MISR and MODIS data, we  
129 calculate regional mean AOD by averaging valid AOD values over all grids within the three  
130 target regions.

131 CALIOP is a dual-wavelength polarization lidar on the CALIPSO satellite, and is  
132 designed to acquire vertical profiles of aerosols and clouds at 532 and 1064 nm wavelengths  
133 during both day and night [Winker et al., 2007]. CALIPSO flies in formation with Aqua, and all  
134 three satellites employed in this paper fly in orbits having 16-day repeat cycles. In addition to  
135 vertical extinction profiles, CALIPSO categorizes an aerosol layer as one of seven types based  
136 on a number of parameters including altitude, location, surface type, volume depolarization ratio,  
137 and integrated attenuated backscatter [Omar et al., 2009]. The seven aerosol types are dust,  
138 smoke, clean continental, polluted continental, polluted dust, clean marine, and dusty marine. For  
139 most profiles, this aerosol classification is consistent with that derived from AERONET  
140 inversion data (Mielonen et al., 2009). In this study, we adopt the Level 2 aerosol profile product  
141 (05kmAPro, V4.10), which has an along-track horizontal resolution of 5 km and a vertical  
142 resolution of 60 m or 180 m, depending on whether the aerosol height is below or above 20.2 km  
143 altitude. We do not use the CALIOP Level 3 product because it is difficult to collocate with  
144 AERONET observations (see Section 2.2) due to its coarse resolution ( $2^\circ \times 5^\circ$ ). For each clear-



145 sky profile, we calculate the column AOD at 532 nm by vertically integrating extinction  
146 coefficients of the features that are identified as “aerosols” and have valid quality control (QC)  
147 flags, i.e.,  $-100 \leq \text{cloud aerosol discrimination (CAD) score} \leq -20$ , extinction QC = 0/1, and  
148 extinction coefficient uncertainty  $< 99.9$  (Huang et al., 2013). In addition, we employ two quality  
149 filters used in generating the Level 3 product in order to eliminate features that probably suffer  
150 from surface contamination, i.e., near-surface features with large negative extinction coefficients  
151 and contaminated features beneath the surface-attached opaque layer (NASA CALIPSO team,  
152 2011). Following the same method, we also bin the 532 nm AODs into various height ranges  
153 above ground level (0-200 m, 200-500 m, 500-800m, 800-1200 m, 1200-2000 m, and  $> 2000$  m  
154 above the surface elevation) for the individual aerosol types. Finally, we derive monthly mean  
155 AODs by averaging all clear-sky aerosol profiles within each month over the three target regions.  
156 Although aerosol extinction coefficients within about 200 m of the surface are considered to be  
157 uncertain despite the application of quality filters (NASA CALIPSO team, 2011), we include  
158 them for completeness but exercise with caution when interpreting variations in AODs  $< 200$  m.  
159 It should be noted that CALIPSO AOD is reported at a different wavelength (532 nm) from those  
160 used in the MISR and MODIS products (555 nm and 550 nm, respectively); this slight  
161 wavelength difference is not expected to affect our conclusions regarding AOD seasonal  
162 variations. Another caveat is that the monthly mean AOD from different sensors is calculated  
163 based on different sets of days, since MODIS provides near-daily global coverage while MISR  
164 and CALIPSO do not. We only use in our analysis monthly AOD averaged across 10 years and  
165 all grids/retrievals within three rather large regions (i.e., EUS, WEU, or ECC). In this case, the  
166 impact of the sampling issue is expected to be much smaller than that on the AOD retrieval in an  
167 individual month at a specific location.



168           2.2 AERONET and surface PM<sub>2.5</sub> data

169           We use AOD observations from AERONET to compare with the AOD seasonal  
170 variations derived from satellite datasets. AERONET sunphotometers directly measure AOD at  
171 seven wavelengths (approximately 340, 380, 440, 500, 675, 870, and 1020 nm) with an estimated  
172 uncertainty of 0.01-0.02 (Holben et al., 2001; Eck et al., 1999), which is much smaller than the  
173 uncertainties associated with satellite measurements (Kahn et al., 2010; Levy et al., 2010;  
174 Schuster et al., 2012). Therefore, we consider AERONET as “ground truth” for AOD temporal  
175 variations. We adopt the AERONET Level 2 Version 2.0 direct-sun measurements of spectral  
176 AODs, which are subsequently interpolated to 550 nm using a second-order polynomial fit to  
177  $\ln(\text{AOD})$  vs.  $\ln(\text{wavelength})$  as recommended by Eck et al. (1999). A fundamental difference  
178 between satellite and AERONET AOD observations is that a satellite acquires data at a single  
179 overpass time (or spread over 7 minutes for MISR’s nine views) and over an extended spatial  
180 area in the case of MISR and MODIS, whereas AERONET obtains a time series of point data at  
181 each surface station. To match coincident measurements, the AERONET AOD retrievals for  
182 each site are averaged within a 2 h window centered on the satellite overpass times (about 10:30  
183 for MISR and MODIS/Terra, and 13:30 for MODIS/Aqua and CALIPSO, depending on site  
184 location), and compared with the satellite AOD retrievals in a  $1^\circ \times 1^\circ$  grid box (consistent with  
185 the grids used in the MODIS Level 3 products) that contains the corresponding AERONET site.  
186 Only those days for which a satellite overpasses an AERONET site are used in the  
187 comparisons. Since AOD variation has a large spatial correlation length of 40-400 km (Anderson  
188 et al., 2003), spatial averaging over a  $1^\circ \times 1^\circ$  grid should not bias the seasonal variations of AOD  
189 but has the benefit of increase the number of data points with valid AOD retrievals that are used  
190 in the comparisons. To assure data quality, only the AERONET sites that span at least 5 years



191 with at least 10 months of valid data in each year are included in the comparison. After screening,  
192 28, 54, and 13 sites are used in our analysis of the EUS, WEU, and ECC regions.

193 To provide additional information on the seasonal variations of satellite-observed aerosol  
194 loadings near the surface, we obtain surface  $PM_{2.5}$  concentrations from several observational  
195 networks over the three target regions. Hourly  $PM_{2.5}$  concentrations for 225 sites over the EUS  
196 region are achieved from the Air Quality System (AQS), which is a large observational database  
197 containing ambient air pollution data collected by the United States Environmental Protection  
198 Agency (USEPA), as well as state, local, and tribal air pollution control agencies in the United  
199 States (USEPA, 2017). For the ECC region, we obtain hourly  $PM_{2.5}$  concentrations from the  
200 Ministry of Environmental Protection of China (MEP, <http://datacenter.mep.gov.cn/>), which  
201 provides continuous measurements at 496 sites located in 74 major cities in China. Hourly/daily  
202  $PM_{2.5}$  concentrations for 52 sites over the WEU region are taken from the European Monitoring  
203 and Evaluation Programme (EMEP). Similar to the processing of AERONET data, we only  
204 include sites whose data span  $\geq 5$  years with  $\geq 10$  months of data in each year, except in the case  
205 of the ECC region where at least 2 years' data are required because the  $PM_{2.5}$  concentrations  
206 have been only publicly available since January 2013.

### 207 **3 Results and Discussion**

#### 208 3.1 Seasonal variations of column AOD

209 Figure 2 illustrates the monthly variations in column AOD observed by MISR,  
210 MODIS/Terra, MODIS/Aqua, and CALIPSO during 2007-2016 in the three target regions. For  
211 consistency with the products from MISR and MODIS, only clear-sky daytime CALIPSO  
212 profiles are used to calculate the lidar-based monthly means. All satellite-borne sensors show  
213 that AOD in the EUS region is the highest in summer and lowest in winter, though CALIPSO



214 reports a noticeably smaller difference between the summer and winter extrema compared with  
215 the other three satellite instruments. For the WEU and ECC regions, MISR, MODIS/Terra, and  
216 MODIS/Aqua also reveal consistent seasonal patterns in which AOD peaks in spring and/or  
217 summer and reaches its lowest valley in winter. CALIPSO, however, shows little intra-annual  
218 variation in AOD, with small peaks occurring in spring and fall.

219 In view of the substantial differences between CALIPSO and the other three sensors, we  
220 compare satellite retrieved AOD seasonal variations with point-based ground measurements  
221 from AERONET (Fig. 3). As in other studies, AERONET data are treated as “ground truth” for  
222 column AOD due to its smaller uncertainty compared with satellite data (Kahn et al., 2010; Levy  
223 et al., 2010; Schuster et al., 2012; Fan et al., 2018). Figure 3 shows that, in all three regions, the  
224 AOD seasonal variations measured by AERONET are similar to those retrieved by MISR,  
225 MODIS/Terra, and MODIS/Aqua, but are quite different from CALIPSO data. Reasons for the  
226 differences between CALIPSO and other sensors will be discussed in Section 3.2. Considering  
227 the high accuracy of AERONET, AOD probably peaks in summer/spring and dips in winter. The  
228 higher AOD in summer is probably explained by accelerated formation of secondary aerosols,  
229 including sulfate, nitrate, ammonium, and secondary organic aerosol (SOA), as a result of  
230 stronger radiation and higher temperature in summer. Another possible reason is the higher RH  
231 in summer which favors the hygroscopic growth of aerosols (Liu et al., 2012; Zheng et al., 2017).

232 While relative pattern of AOD seasonal variations from observations of MISR,  
233 MODIS/Terra, and MODIS/Aqua are similar to each other and to those of AERONET, the  
234 magnitude of AOD observed by these sensors shows remarkable discrepancies. The MISR-  
235 retrieved AOD agrees well with the AERONET observations in EUS and WEU regions (Fig. 3).  
236 In the ECC region, however, MISR underestimates the AERONET AOD, probably because there



237 is less signal from the surface at higher AOD, which creates ambiguity that can result in the  
238 algorithm assigning too much of the top-of-atmosphere radiance to the surface (i.e., a higher  
239 surface albedo), thereby underestimating the AOD (Kahn et al., 2010). The MODIS/Terra and  
240 MODIS/Aqua significantly overestimate the AERONET AOD in EUS and WEU regions, and  
241 slightly overestimate the AERONET AOD in the ECC region, which is consistent with the  
242 evaluation results of Remer et al. (2005). This overestimation is largely attributed to the  
243 systematic positive bias at low AOD, likely attributed to an instrument calibration issue or an  
244 improper representation of surface reflectance at certain locations and seasons (Remer et al.,  
245 2005). The much smaller overestimation in the ECC region is explained by the fact that the  
246 MODIS AOD is not overestimated or even underestimated at the high AOD range, probably due  
247 to insufficient light absorption in the aerosol models (Remer et al., 2005).

### 248 3.2 Seasonal variations of aerosol loadings as a function of height

249 In addition to column AOD, the climatic effects of aerosols are also strongly dependent  
250 on their vertical distribution. To explore intra-annual variations in aerosol vertical profile, Fig. 4  
251 presents CALIPSO-observed monthly variations of AOD as a function of height in the three  
252 target regions. A striking pattern is that the AOD seasonal variations are dramatically different at  
253 lower and upper heights. Over the WEU and ECC regions, AODs of the vertical layers below  
254 800 m generally peak in winter, while those above 800 m peak in summer/spring. As a result, the  
255 CALIPSO-observed column AOD for these two regions presents a rather uniform seasonal  
256 pattern. For the EUS region, the maximum AOD above 800 m also occurs in summer; however,  
257 AOD below 800 m shows two peaks, one in summer and the other in winter. The integration of  
258 various layers thus yields a nearly unimodal distribution with maximum occurring in summer.



259 To provide an independent evaluation of the CALIPSO-observed AOD variations at  
260 lower heights, we examine the seasonal variations of near-surface PM<sub>2.5</sub> concentrations at  
261 hundreds of surface monitor locations within the three target regions. The aerosol extinction  
262 coefficient, and hence AOD at lower heights is affected by not only the particle mass  
263 concentrations, but also aerosol type (absorbing vs. nonabsorbing aerosols, coarse-mode vs. fine-  
264 mode aerosols) and meteorological parameters such as RH, wind speed and direction, and  
265 planetary boundary layer height (Zheng et al., 2017). Nevertheless, previous studies have  
266 reported fairly good correlations between extinction coefficient/low level AOD and PM<sub>2.5</sub>  
267 concentrations (Cheng et al., 2013; Zheng et al., 2017). For this reason, it is reasonable to  
268 qualitatively compare the seasonal variation patterns of near-surface PM<sub>2.5</sub> concentrations and  
269 low-level AOD. Figure 5 shows that, over the ECC and WEU regions, surface PM<sub>2.5</sub>  
270 concentrations are largest in winter and smallest in summer. In the EUS region, the maximum  
271 PM<sub>2.5</sub> concentration occurs in summer and a second maximum occurs in winter. These trends are  
272 generally consistent with the seasonal variations of AOD at low heights, implying that CALIPSO  
273 data can generally capture the seasonal changes in low-level aerosol abundance.

274 The aerosol vertical distribution is an important factor in reconciling CALIPSO and other  
275 sensors with regard to AOD seasonal variations. MISR, MODIS, and AERONET all measure  
276 column-integrated AOD using spectroradiometers, whereas CALIOP is an active lidar which  
277 estimates vertically-resolved AOD based on vertical profiles of attenuated backscatter. By  
278 comparing CALIPSO with the Atmospheric Radiation Measurement (ARM) program's ground-  
279 based Raman lidars, Thorsen et al. (2017) showed that CALIPSO does not detect all relatively  
280 significant aerosols due to insufficient detection sensitivity, and that the fraction of aerosols  
281 detected in the upper air is much smaller than that near the ground surface because the upper-



282 level aerosols tend to be optically thin. Therefore, the CALIPSO-observed AOD seasonal  
283 variations are significantly weighted toward lower heights. Specifically, over WEU and ECC  
284 regions, the unimodal AOD distributions with a summer peak at higher levels are largely  
285 counteracted by the opposite seasonal variations at lower levels, resulting in rather uniform  
286 seasonal variations of column AOD. For the EUS regions, due to the bimodal AOD distribution  
287 at lower heights, the summer peak in column AOD variations remain but the difference between  
288 peak and valley is smaller than implied by the observations of MISR and MODIS. In this sense,  
289 although the integrated CALIPSO column AOD does not agree well with AERONET, it does  
290 provide valuable information with respect to intra-annual variations of AOD at specific height  
291 ranges.

292         Why are the AOD seasonal variations different between the lower and upper atmosphere?  
293 The atmosphere in winter is generally more stable and vertical mixing is weaker, therefore more  
294 aerosols, particularly primary aerosols, are confined to lower heights, resulting in the peak of  
295 low-level AOD in winter (Guo et al., 2016; Liu et al., 2012; Zheng et al., 2017). At higher levels,  
296 the maximum AOD in summer can be explained by two reasons: (1) more aerosols, especially  
297 primary aerosols, are transported to the upper level in summer due to stronger vertical mixing  
298 (Guo et al., 2016; Liu et al., 2012; Zheng et al., 2017), and (2) secondary aerosol formation is  
299 more rapid in summer because of stronger radiation and higher temperature, and much of the  
300 secondary aerosols are produced in the upper air (de Reus et al., 2000; Minguillon et al., 2015;  
301 Heald et al., 2005). In addition, the seasonal variations of AOD at different vertical levels may  
302 also be influenced by the variations of RH which affects the hygroscopic growth (Liu et al., 2012;  
303 Zheng et al., 2017) as well as the seasonal patterns of inter-regional transport of aerosols (Tian et  
304 al., 2017).



## 305 3.3 Seasonal variations of aerosol types

306 Besides column AOD and vertical profiles, another factor influencing aerosol climate  
307 impact is aerosol type (i.e., partitioning by size and chemical composition). The MISR and  
308 CALIPSO products classify aerosols based on distinct principles of measurement and retrieval  
309 algorithms. Analysis of the two datasets in combination can potentially lead to a deeper  
310 understanding of the factors driving temporal variations of aerosol type. Figures 6 and 7 illustrate  
311 the seasonal variations of AODs for various aerosol types retrieved by MISR and CALIPSO,  
312 respectively. As discussed in Section 3.2, relative variability in CALIPSO-derived AOD at  
313 different height ranges appears to be more reliable than integrated column AOD, therefore we  
314 show aerosol types below and above 800 m separately in Fig. 7.

315 MISR distributes AODs into three size ranges, i.e., small ( $< 0.7 \mu\text{m}$  diameter), medium  
316 ( $0.7\text{-}1.4 \mu\text{m}$  diameter), and large ( $> 1.4 \mu\text{m}$  diameter). Among the major constituents of ambient  
317 aerosols, which include primary aerosols (dust, sea-spray aerosols, and primary anthropogenic  
318 aerosols) and secondary aerosols (sulfate, nitrate, ammonium, and SOA), dust and sea-spray  
319 aerosols are predominantly coarse particles and secondary aerosols are dominated by very fine  
320 particles, while primary anthropogenic aerosols span a large size range, leading to a mean size  
321 intermediate between dust/sea-spray and secondary constituents (Seinfeld and Pandis, 2006). Fig.  
322 6 indicates that the small-size AOD is much larger in spring/summer than in winter over all  
323 regions, whereas large-size AOD generally shows rather uniform distributions, except for the  
324 ECC region where a peak occurs in late winter/early spring. The high small-size AOD in summer  
325 is probably due to accelerated secondary aerosol formation and enhanced hygroscopic growth, as  
326 described in Section 3.1. In contrast, AOD of primary anthropogenic aerosols should be less  
327 influenced by seasonal effects, which partly accounts for the rather uniform distributions of



328 large-size AOD. Additionally, the changes in large-size AOD are also affected by other aerosol  
329 components including dust and sea-spray aerosols, as discussed below.

330 In contrast to MISR's partitioning of aerosol type by size, shape, and absorption, the  
331 CALIPSO-retrieved aerosol types (Fig. 7) are characterized by emission source. Particles  
332 associated with anthropogenic air pollution (polluted continental and polluted dust) comprise the  
333 dominant type in all three regions. At higher levels, the maximum AOD of polluted  
334 continental/dust aerosols occurs in spring/summer, while the maximum occurs in winter at lower  
335 levels (plus a second maximum in summer in EUS), in accordance with the seasonal variations  
336 of total AOD at different heights, as discussed in Section 3.2. With regard to dust and clean  
337 marine (sea-spray) aerosols, the AOD in the EUS region does not show an obvious seasonal  
338 pattern. In the WEU region, AOD of dust aerosols peaks in summer, consistent with previous  
339 surface-based observational studies which show that dust events in Europe predominantly occur  
340 during summer due to transport from the Sahara region (Stafoggia et al., 2016). The AOD of dust  
341 above 800 m is much larger than that below 800 m, supporting the conclusion that dust aerosols  
342 in WEU mainly originate from long range transport. Since the dust AOD has a quite large inter-  
343 annual variation (denoted by the error bars in Fig. 7), we use the Student's t-test to demonstrate  
344 the statistical significance of the seasonal variations shown above. The dust AOD in summer is  
345 statistically larger than that in any other season at the 0.05 level, indicating the robustness of the  
346 peak in summer. Contrary to dust, the AOD of sea-spray aerosols in WEU is much higher in  
347 winter than in summer, probably because winter is the relative windy season with large low  
348 pressure systems over the Atlantic Ocean and the North Sea (Manders et al., 2009). The offset of  
349 the opposite variation trends in dust and sea-spray aerosols partly accounts for the rather uniform  
350 distributions of large-size AOD (see Fig. 6). Over the ECC region, sea-spray aerosols make a





351 negligible contribution to total AOD. The dust AOD is much larger in spring than in any other  
352 season (significant at the 0.05 level), which is tied to the outburst of springtime Gobi desert dust  
353 storms (China Meteorological Administration, 2012). The high dust AOD explains the peak in  
354 large-size AOD in spring over the ECC region (see Fig. 6).

355 Smoke aerosols are predominantly located above 800 m. Over the EUS and WEU regions,  
356 smoke aerosols present a unimodal distribution with maximum occurring in summer. The  
357 differences between smoke AOD in summer and the other three seasons are all statistically  
358 significant at the 0.05 level, except for the difference between summer and spring over the WEU  
359 region, which is statistically significant at the 0.10 level. In the ECC region, the smoke AOD  
360 follows a bimodal distribution with peaks occurring in March and August and valleys occurring  
361 in May and December. The differences between either of the peak months and either of the  
362 valley months are statistically significant at the 0.05 level. MISR's independent retrieval of  
363 absorbing AOD (Fig. 6) presents a similar seasonal pattern (statistically significant at the 0.05  
364 level) as the CALIPSO smoke AOD. In fact, smoke consists of a larger fraction of absorbing  
365 aerosols (Dubovik et al., 2002), such as BC and light-absorbing organic aerosol (Kirchstetter and  
366 Thatcher, 2012), as compared to other aerosol types. The variability of MISR absorbing AOD  
367 (shown in the right Y-axis of Fig. 6) is about 0.002-0.005, while the variability of smoke AOD  
368 from CALIPSO is about 0.01-0.03. The smoke AOD includes the contributions of both the  
369 absorbing and scattering portions. The MISR absorbing AOD, which is calculated using total  
370  $AOD \times (1 - \text{single scattering albedo})$ , represents only the absorbing portion but includes  
371 contributions from aerosol types other than smoke (Bull et al., 2011). Considering that the single  
372 scattering albedo of smoke is about 0.80-0.94 (Dubovik et al., 2002), the MISR absorbing AOD  
373 and CALIPSO smoke AOD are consistent in the order of magnitude. For the preceding reasons,



374 the seasonal patterns of smoke and absorbing aerosols acts as a cross-validation and strengthens  
375 the reliability of the observed trends. Over the EUS and WEU regions, the largest smoke AOD in  
376 summer could be explained by the highest emissions from forest and grassland fires (van der  
377 Werf et al., 2017). Over the ECC region, an additional peak occurs in March because agricultural  
378 residue burning makes a substantial contribution to total smoke emissions (van der Werf et al.,  
379 2017), and such burning takes place more frequently in March due to burning of crop residues  
380 left on the fields from the previous growing season (Shon, 2015).

#### 381 3.4 Diurnal variations of height- and type-resolved AODs

382 CALIPSO provides both daytime and nighttime aerosol retrievals, giving an opportunity  
383 to analyze aerosol diurnal variations. Care must be exercised in interpreting the CALIPSO-  
384 observed diurnal variability because CALIPSO's detection sensitivity is lower during daytime  
385 due to interference from sunlight. As a result, the difference between daytime and nighttime  
386 AODs represents an overall effect of both actual and artificial diurnal variability. In this section,  
387 we identify variation patterns that are most likely real, in spite of the possible effects of  
388 measurement artifacts.

389 Figure 8 shows the differences between nighttime and daytime total and height-resolved  
390 AODs (nighttime minus daytime values). The sign of the difference depends on the relative  
391 importance of two competing factors: a more stable atmosphere at night favors the accumulation  
392 of aerosols, whereas stronger photochemical reactions enhance aerosol loading during the day.  
393 Fig. 8 reveals that nighttime AODs tend to be smaller than daytime AODs in winter. In contrast,  
394 during summer and/or spring, nighttime AODs are higher than (WEU and ECC) or similar to  
395 (EUS) the daytime values. More precisely, the difference between nighttime and daytime AODs  
396 is significantly more positive in summer. This pattern is applicable to both AODs < 800 m and >



397 800 m. It is unlikely to be attributable to the difference in CALIPSO's detection sensitivity,  
398 which presumably exerts similar effects in all seasons. It should be noted, however, that we  
399 cannot fully exclude the effects of measurement artifacts, the quantification of which require  
400 further investigations using more accurate measurement techniques. A probable reason for more  
401 positive nighttime-daytime AOD difference in summer is that the diurnal temperature range,  
402 which is defined as the difference between the maximum and minimum temperatures in a day, is  
403 larger in summer (Ruschy et al., 1991; Jackson and Forster, 2010; Sun et al., 2006), giving rise to  
404 a larger difference in nighttime and daytime AODs due to a more stable nighttime atmosphere  
405 compared to a more unstable daytime atmosphere.

406 We further illustrate the differences between nighttime and daytime AODs of major  
407 aerosol types (Fig. 9). Smoke AOD is much higher in the nighttime than in the daytime.  
408 Considering that the emission rates of smoke aerosols is likely to be similar at different time of  
409 the day, the higher nighttime smoke AOD is probably the result of increased atmospheric  
410 stability at night, allowing the aerosols to accumulate. On the contrary, dust AOD is usually  
411 higher in the daytime, which may be tied to higher near-surface wind speed in the day (He et al.,  
412 2013; He et al., 2012; Hasson et al., 1990). The diurnal variations of polluted continental/dust  
413 aerosols vary according to season and height, and likely depend on the relative roles of more  
414 stable atmosphere at night and more active chemical reactions in the day. The type-dependent  
415 diurnal variations should mainly be representative of actual conditions, as such different  
416 variations as a function of aerosols type are unlikely explained by instrument detection  
417 sensitivity.



#### 418 **4 Conclusions and implications**

419 This study investigated the seasonal and diurnal variations of aerosol column loading,  
420 vertical distribution, and particle types using multiple satellite and ground-based observational  
421 datasets during 2007-2016 over EUS, WEU, and ECC regions. Retrievals from MISR and  
422 MODIS reveal that column AOD in all three regions peaks in spring/summer and reaches its low  
423 in winter, which is consistent with observations from AERONET. This seasonal pattern is  
424 probably explained by accelerated formation of secondary aerosols in spring/summer due to  
425 stronger insolation and higher temperature. In contrast, CALIPSO shows a much weaker  
426 seasonal variability in column AOD, probably because CALIPSO-retrieved AOD is weighted  
427 toward lower heights since some thin aerosol layers in high levels are undetected due to  
428 insufficient detection sensitivity. Despite the discrepancy in integrated column AOD, CALIPSO  
429 does provide valuable information with respect to intra-annual variations of AOD as a function  
430 of height. Over the WEU and ECC regions, AODs of the vertical layers below 800 m generally  
431 peak in winter, while those above 800 m mostly peak in summer. For the EUS region, the  
432 maximum AOD above 800 m also occurs in summer; however, AOD below 800 m shows two  
433 peaks, one in summer and the other in winter. The seasonal variations of AOD at low heights are  
434 consistent with seasonal patterns of measured surface PM<sub>2.5</sub> concentrations.

435 When aerosols are binned into different size ranges, the small-size AOD is much larger in  
436 spring/summer than in winter over all three regions. Large-size AOD generally shows rather  
437 uniform distributions, except for the ECC region where a peak occurs in spring, consistent with  
438 the largest dust AOD in this season. When aerosols are classified according to sources, the  
439 aerosols associated with anthropogenic air pollution (as well as mixtures of anthropogenic  
440 pollution and dust) are the dominant type in all three regions. AOD of polluted aerosols has a



441 similar seasonal pattern as total AOD. Dust and clean marine aerosols in the WEU region peak in  
442 summer and winter, respectively, whereas they do not show an obvious seasonal pattern in the  
443 EUS region. Smoke aerosols, which CALIPSO indicates are predominantly located at heights  
444 above 800 m, present an obvious unimodal distribution with maximum occurring in summer over  
445 EUS and WEU regions, and a bimodal distribution with peaks in August and March over the  
446 ECC region. This pattern is in good agreement with the seasonal variations of absorbing AOD  
447 derived from MISR.

448       Regarding diurnal variations, the difference between nighttime and daytime AODs  
449 (nighttime minus daytime) is more positive in summer than in winter, which is likely explained  
450 by larger diurnal temperature range in summer. Smoke AOD is much higher in the nighttime,  
451 when the atmosphere is more stable, than in the daytime. On the contrary, dust AOD is usually  
452 higher in the daytime, when higher winds speeds enable a greater abundance of particles to  
453 become airborne.

454       The combination of multiple satellite and ground-based observations facilitate a  
455 systematic and deeper understanding of the seasonal and diurnal variations of aerosols,  
456 particularly their vertical and type distribution. Comparison of multiple measurement and  
457 retrieval methodologies enables reducing the uncertainties in the estimation of aerosol direct  
458 effects by providing improved information about aerosol vertical and type distributions, which  
459 significantly affect the aerosol-induced scattering and absorption of radiation. More importantly,  
460 the intra-annual variations of vertical distributions and types of aerosols are important for  
461 understanding their impact on atmospheric dynamics, cloud fields, and precipitation production.  
462 Many studies (Chen et al., 2017; Bond et al., 2013) have shown that BC and dust can either  
463 enhance or inhibit convection and hence cloud fields, depending on their vertical locations,



464 which are very different from the effects of non-absorbing aerosols (Ramanathan et al., 2005;  
465 Fan et al., 2008; Massie et al., 2016). Finally, the data and variation patterns presented in this  
466 study can be used to evaluate and improve model simulations, with the ultimate goal of  
467 improving model assessment of the climatic and health effects of aerosols.

468

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475

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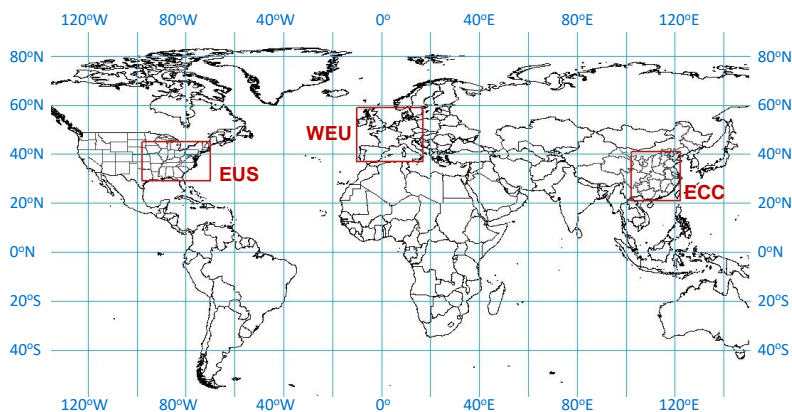
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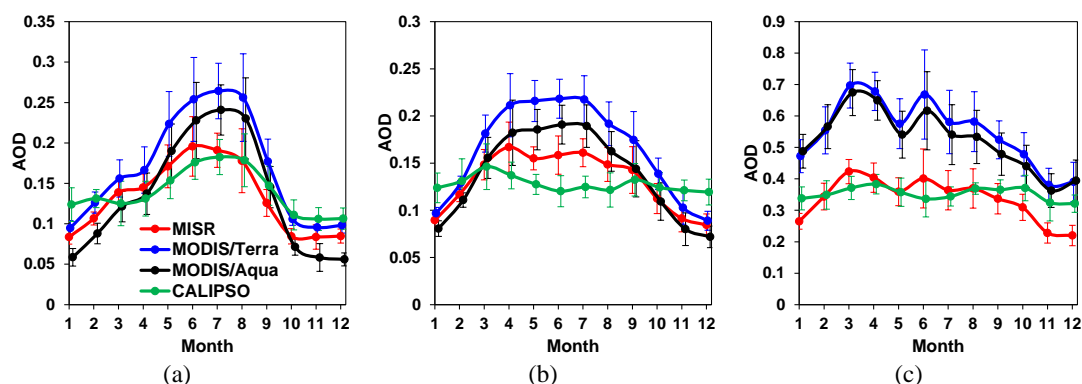


## 699 Figures

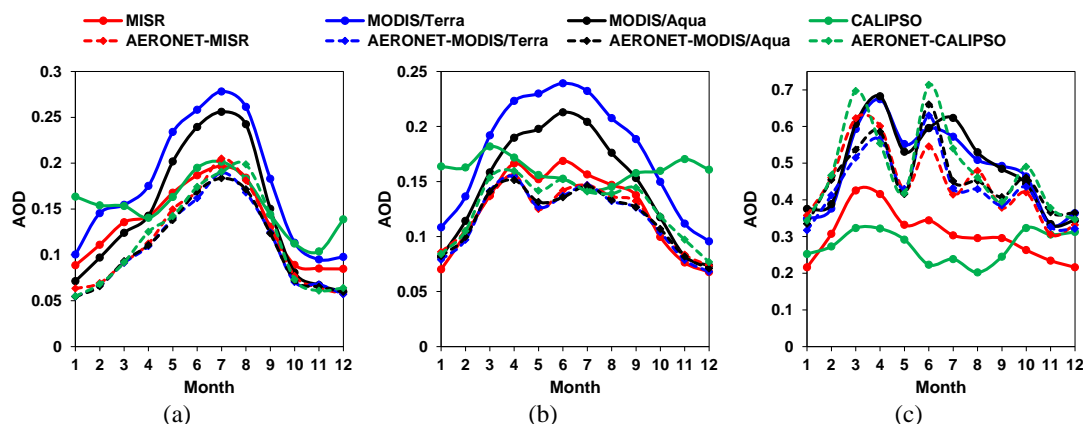


700  
701 **Figure 1.** Target regions for this study: the Eastern United States (EUS), Western Europe  
702 (WEU), and Eastern and Central China (ECC).

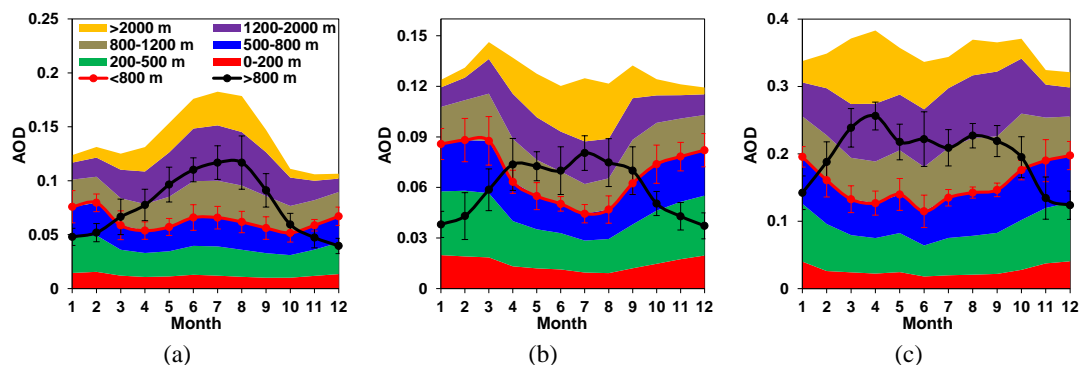
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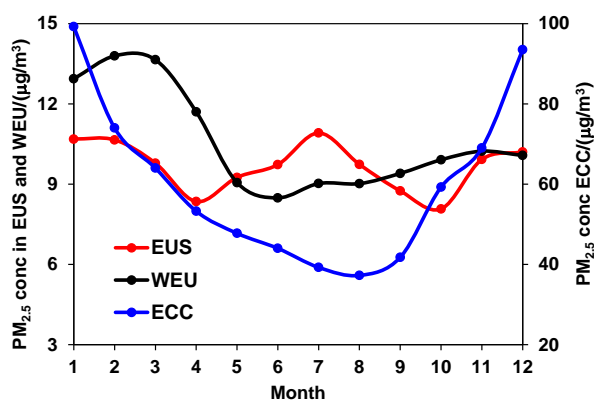
704 **Figure 2.** Monthly mean AOD observed by MISR, MODIS/Terra, MODIS/Aqua, and CALIPSO  
705 during 2007-2016 in (a) EUS, (b) WEU, and (c) ECC. For CALIPSO, only clear-sky daytime  
706 profiles are averaged in order to be consistent with the MISR and MODIS products. The error  
707 bars denote the standard deviation of the monthly mean AOD values obtained over all years.  
708 Note the different scales on the y-axes of the plots.



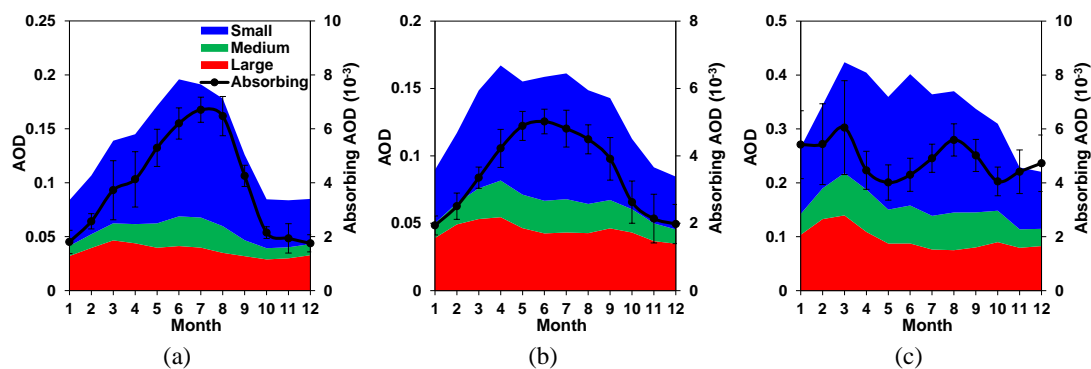
709 **Figure 3.** Monthly mean AOD observed by satellites and AERONET averaged across the  
710 AERONET sites during 2007-2016 in (a) EUS, (b) WEU, and (c) ECC. The observations from  
711 MISR, MODIS/Terra, MODIS/Aqua, and CALIPSO are averaged over  $1^{\circ}\times 1^{\circ}$  grid boxes  
712 containing the AERONET sites. The AERONET data are averaged within a 2 h window  
713 centered on satellite overpass times. The numbers of AERONET sites included in analysis are  
714 28, 54, and 13, in the EUS, WEU, and ECC regions, respectively. Since the four sensors  
715 overpass a site in different days and different times of day, we separately calculate the  
716 AERONET data matched to each sensor (denoted by “AERONET- $\times\times\times$ ”). The AERONET curves  
717 matched to different sensors are close in EUS and WEU, partly because there are plenty of  
718 sites in these two regions, and the discrepancy due to the sampling issue is therefore smoothed  
719 out. In contrast, there are only 13 AERONET sites in ECC, so there exists larger discrepancy  
720 between the AERONET data matched to different sensors. Note the different scales on the y-  
721 axes of the plots.



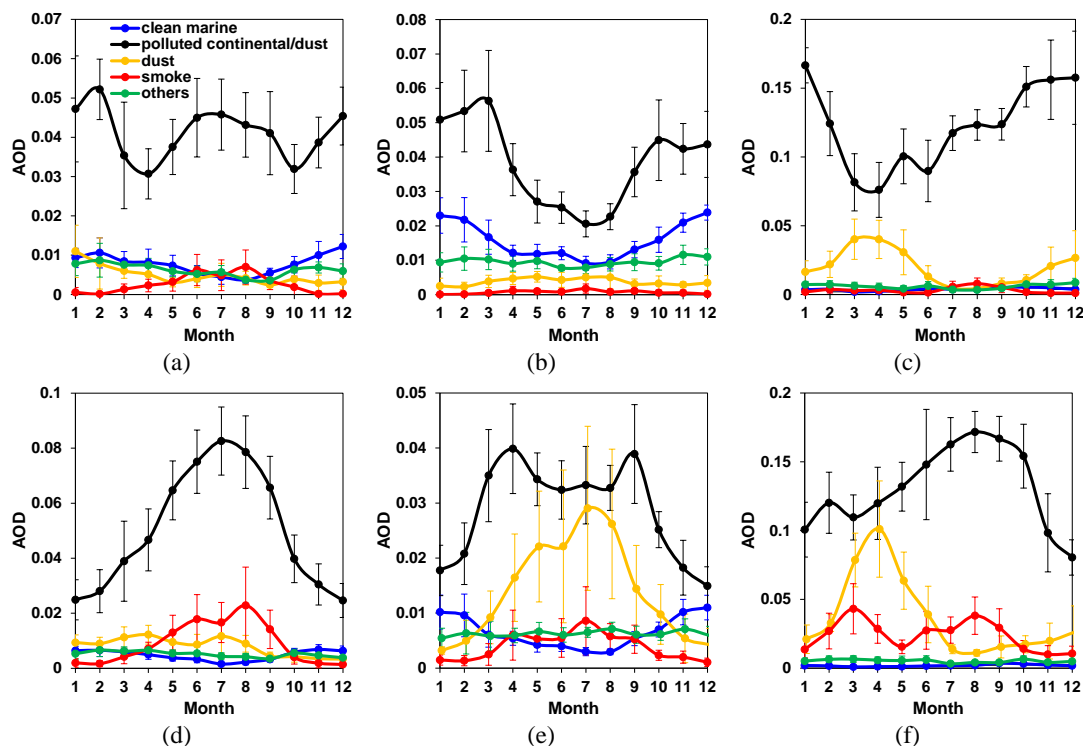
722 **Figure 4.** Monthly mean AOD as a function of height above ground level observed by CALIPSO  
 723 during 2007-2016 in (a) EUS, (b) WEU, and (c) ECC. Only clear-sky daytime profiles are  
 724 averaged in order to be consistent with the products of MISR and MODIS. The range of AOD  
 725 within a particular height range is depicted by the colored stacks. The integrated AODs for  
 726 heights below and above 800 m are shown as solid lines, for which the error bars are defined in  
 727 the same way as in Fig. 2. Note the different scales on the y-axes of the plots.



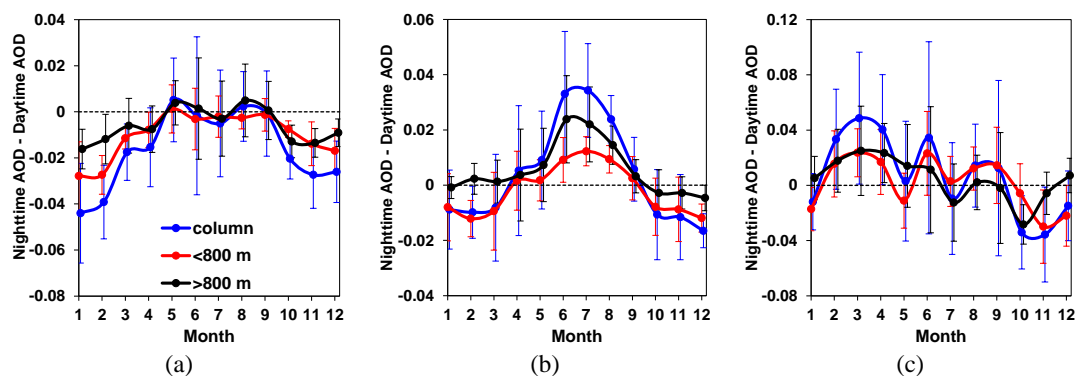
728 **Figure 5.** Monthly mean surface  $PM_{2.5}$  concentrations during 2007-2016 in three target regions.  
 729 The numbers of observational sites included in averaging are 225, 52, and 496, in the EUS,  
 730 WEU, and ECC regions. Note the different scales on the y-axes for EUS/WEU and ECC.  
 731



732 **Figure 6.** Monthly mean AOD of different aerosol types observed by MISR during 2007-2016 in  
733 (a) EUS, (b) WEU, and (c) ECC. The size-resolved AODs are depicted by the colored stacks  
734 (left Y-axis); the integration of the three size ranges yields total column AOD, as represented by  
735 the upper edge of the blue color. The AOD of absorbing aerosols is shown as solid lines (right  
736 Y-axis), for which the error bars are defined in the same way as in Fig. 2. Note the different  
737 scales on the y-axes of the plots.

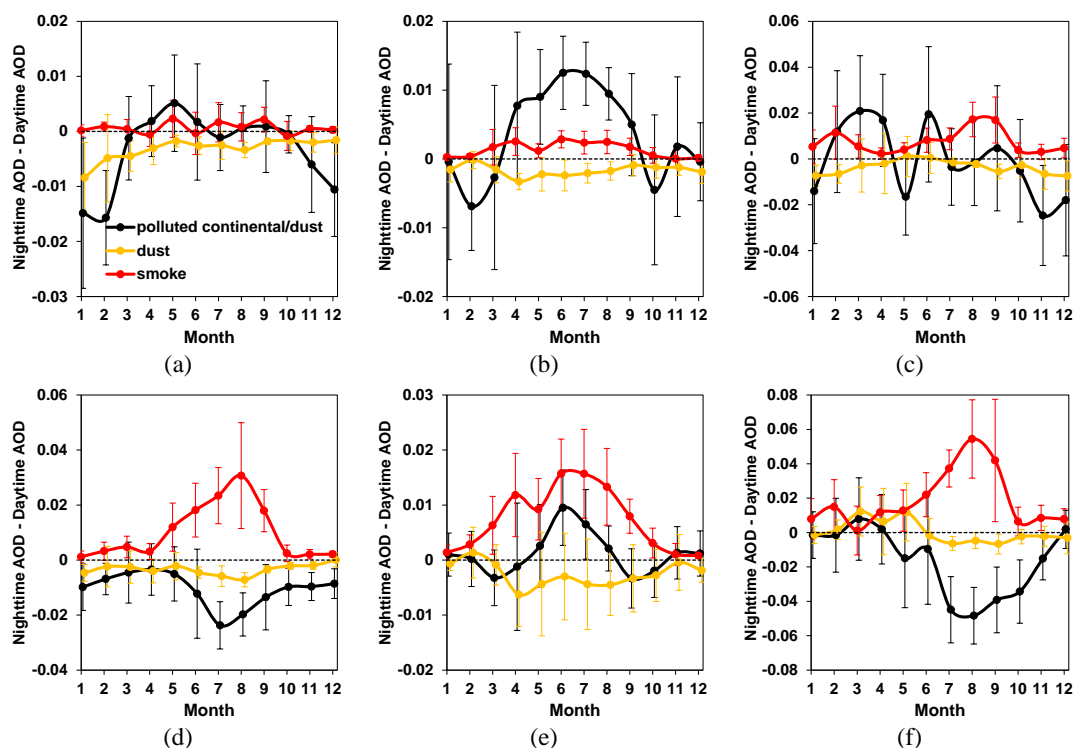


738 **Figure 7.** Monthly mean AOD of different aerosol types (a-c) below 800 m and (d-f) above  
739 800 m observed by CALIPSO during 2007-2016 in (a, d) EUS, (b, e) WEU, and (c, f) ECC. Only  
740 clear-sky daytime profiles are used in the averaging to be consistent with the products of MISR  
741 and MODIS. The definition of error bars is the same as in Fig. 2. Note the different scales on the  
742 y-axes of the plots.



743 **Figure 8.** Differences between monthly mean nighttime and daytime AODs (including column  
744 AODs, as well as AODs of aerosols located below 800 m and above 800 m) observed by  
745 CALIPSO during 2007-2016 in (a) EUS, (b) WEU, and (c) ECC. Only clear-sky profiles are  
746 included. The definition of error bars is the same as in Fig. 2. Note the different scales on the y-  
747 axes of the plots.





748 **Figure 9.** Differences between monthly mean nighttime and daytime AODs of different aerosol  
749 types (a-c) below 800 m and (d-f) above 800 m observed by CALIPSO during 2007-2016 in (a,  
750 d) EUS, (b, e) WEU, and (c, f) ECC. Only clear-sky profiles are included. The definition of error  
751 bars is the same as in Fig. 2. Note the different scales on the y-axes of the plots.

752