1	Multi-angle aerosol optical depth retrieval method based on
2	improved surface reflectance
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12	Abstract
13	Retrieval of atmospheric aerosol optical depth (AOD) has been a challenge for Earth satellite

14 observations, mainly due to the difficulty in estimating surface reflectance with the combined 15 influence of land-atmosphere coupling. Current major satellite AOD retrieval products have low spatial resolution under complex surface processes. In this study, we further improved the surface 16 17 reflectance by modeling the error correction based on the previous AOD retrieval and obtained 18 more accurate AOD retrieval results. A lookup table was constructed using the Second Simulation 19 of Satellite Signal in the Solar Spectrum (6S) to enable high-precision AOD retrieval. The 20 accuracy of the algorithm's retrieval was verified by the Aerosol Robotic Network observations 21 (AERONET). From the validation results, we find that among the nine Multi-angle Imaging

22	SpectroRadiometer (MISR) angles, the retrieved AOD has the best retrieved results with the AOD
23	observed at the An angle (Taihu: R=0.81, RMB=0.68; Xuzhou-CUMT: R=0.73, RMB=0.78). This
24	study will help to further improve the retrieval accuracy of multi-angle AOD at large spatial scales
25	and long time series. The retrieved AOD based on the improved method has the advantages of
26	fewer missing pixels and finer spatial resolution compared to the MODIS AOD products and our
27	previous estimates.
28	Keywords: surface reflectance; aerosol optical depth; satellite remote sensing; MISR; MODIS
29	1. Introduction
30	Aerosols are liquid or solid particles suspended in the atmosphere, with particle diameters
31	ranging from approximately 0.001 to 100 µm (Giles et al., 2019). Aerosol particles suspended in

32 the atmosphere not only affect the radiative forcing of the Earth's air system through direct, 33 indirect or semi-direct action, but also influence climate, and cause environmental problems such as acid rain and haze (Hatzianastassiou et al., 2009; Dao et al., 2014; Daniel et al., 2014; Samset et 34 35 al., 2018; Li et al., 2018; Huang et al., 2021). Therefore, the study of aerosol distribution and 36 change trend has become a hot topic in the field of aerosol research (Holben et al., 2001; Li et al., 37 2020; Berhane et al., 2021; Sun et al., 2022). In addition, high concentrations of aerosols can pose 38 a serious threat to human health (Lee et al., 2010; Dehghani et al., 2012; Mironova et al., 2015). 39 The optical properties of aerosols include parameters such as aerosol optical depth (AOD), 40 scattering phase function, single scattering albedo, and absorbing optical depth. Among them, 41 AOD is an important parameter defined as the integral of the aerosol extinction coefficient in the 42 vertical direction. It describes the attenuation effect of aerosols on light and is an important 43 indicator of the level of air pollution. Multi-channel spectrometers on board geostationary and

44 polar orbiting satellites have been used for AOD retrieval in the last two decades. The AOD 45 products obtained through satellite retrievals are widely used in atmospheric environmental 46 studies (Kaufman et al, 1997; Xie et al., 2019; Chen et al., 2021). Although the accuracy of AOD 47 retrieval has been improving, there is still much room for improvement in the AOD retrieval on 48 land when more complex influences are taken into account.

49 Scholars have conducted research on AOD retrieval using satellite-based multi-angle sensors. Kokhanovsky et al. (2009) used the Along-Track Scanning Radiometer Dual View (ATSR-DV) 50 51 algorithm to retrieve AOD over Germany and compared the retrieval results with the Medium 52 Resolution Imaging Spectrometer (MERIS) and Multi-angle Imaging SpectroRadiometer (MISR) products. The results show that the ATSR-2 algorithm is also applicable to the Advanced 53 54 Along-Track Scanning Radiometer (AATSR). Sundstrom et al. (2012) obtained an aerosol model 55 of eastern China based on Aerosol Robotic Network (AERONET) observation data and used the 56 ATSR-DV algorithm to retrieve the proportion of AOD and coarse to fine particles from AATSR 57 data. Abdou et al. (2005) compared the MISR AOD and the Moderate-resolution Imaging 58 Spectroradiometer (MODIS) AOD products carried by Terra using data from 62 AERONET 59 observation sites. The results show that over land, the MODIS AOD of 0.470 um and 0.660 um 60 channels is 35% and 10% higher than that of MISR, respectively. In coastal and desert areas, MODIS retrieval error is large, while over the ocean, at 0.470 um and 0.660 um channels, MISR is 61 62 0.1 and 0.05 higher than MODIS AOD value, mainly depending on the accuracy of radiometric 63 calibration. Martochik et al. (1997) proposed an algorithm for extracting aerosol optical parameters using MISR multi-angle observations. The results show that the low reflectance of 64 vegetation and multi-angle observation data are used to retrieve AOD in the case of dense land 65

vegetation. As a new remote sensing tool, multi-angle remote sensing can provide aerosol optical depth, single scattering albedo, phase function and other features with sufficient accuracy, which is more suitable for playing its unique role in aerosol research than traditional single-angle optical remote sensing (Dubovik et al., 2019). Multi-angle remote sensing retrieval of aerosol optical properties can use the angle information contained in satellite signals to better separate the contributions of the surface and atmosphere, which is suitable for some bright surfaces. This provides a new way for AOD retrieval.

73 Surface reflectance measures the ability of land to absorb and reflect solar radiation. On land, 74 surface reflectance is relatively complex and is detected by satellite sensors after atmospheric 75 scattering and absorption. Satellite observations capture the combination of there two components, 76 making it challenging to directly separate surface reflectance from atmospheric scattering. 77 Therefore, simultaneous acquisition of atmospheric aerosols and surface reflectance is the main objective of quantitative satellite remote sensing (Deuzé et al., 2001). In optical remote sensing, 78 79 the blue wavelength is shorter and the surface reflectance is lower, resulting in more atmospheric 80 reflection and scatteringm. Blue band is commonly used for AOD retrieval. In the process of AOD 81 retrieval, overestimation of surface reflectance leads to underestimation of AOD, while 82 underestimation of surface reflectance leads to overestimation of AOD. In general, aerosol signals are weaker than surface signals (Dong et al., 2023). Previous studies have shown that when the 83 84 surface reflectance intercept error is 0.01, the error of satellite remote sensing retrieval AOD is 85 about 0.1 (Zhang et al., 2021). Therefore, accurate estimation of surface reflectance is very 86 importance for the accurate retrieval of aerosol optical depth.

87

In order to improve the accuracy and resolution of MISR AOD retrieval, we analyzed the

MISR AOD retrieval errors of 9 camera angles in previous studies. Secondly, by establishing an error correction model, we modify the surface reflectance of MISR to improve the accuracy of surface reflectance estimation. Finally, the improved AOD is verified with the ground observation site AOD data, and the retrieval results are compared with the previous retrieval results and MODIS AOD products.

93 2. MISR, MODIS, and AERONET Data

94 **2.1 MISR data**

95 In this study, we employed MISR Level 1B2 terrain data (MI1B2T) and used HEGTool 96 (HDF-EOS To GeoTIFF Conversion Tool) software to extract radiation data. Blocks 64 and 65 97 covering the Yangtze River Delta region were extracted by selecting corresponding blocks, output 98 data types and projections based on the regional locations of 180 blocks (Figure S1). In addition, 99 we extracted the solar zenith angle, solar azimuth angle, satellite zenith angle, and satellite azimuth angle data of 9 cameras from the angle dataset (MI1B2GEOP) and selected the 100 101 corresponding blocks, output data types and projection information. To reduce the influence of 102 cloud, we carried out cloud detection and cloud pixel removal on satellite remote sensing images 103 with cloud pixel coverage less than 50%, and used blue band threshold to remove cloud pixels. 104 However, through repeated experiments, we found that the fixed threshold was not effective in 105 removing cloud pixels at 9 angles of the MISR sensor (Figures S2 and S3). Therefore, we use 106 dynamic threshold method to remove cloud pixels from MISR data. Details of the data used in this 107 study are shown in Table S2.

108 2.2 MODIS data

109 In this study, we used MODIS L1B data, including radiance data (MOD02/MYD02) and geolocation data (MOD03/MYD03). Data preprocessing using the MODIS Conversion Toolkit 110 111 (MCTK) tool, including radiation calibration, butterfly processing, geometric correction, reprojection, and band extraction. MODIS BRDF/Albedo is a standard Level 3 product that 112 represents the surface characteristics of MODIS instruments on the Terra and Aqua satellite 113 114 platforms. The product retrieval cycle is 16 days, and the observations on day 9 of each 16-day 115 cycle are weighted to generate daily data, called the global daily surface albedo product (Hsu et al., 2004). The core data set for the MODIS BRDF product is MCD43A1. 116

117 **2.3 AERONET data**

118 AERONET uses the CE-318 instrument, a solar radiation photometer produced in France, 119 which obtains direct solar spectral radiation measurements at 340, 380, 440, 500, 670, 870, 936, 1020 and 1640 nm channels at 1.5° field angles every 3 min. The total water vapor content of the 120 121 atmosphere can be obtained in the 936 nm channel. The AOD value can be retrieved by the 122 measured value of other channels, and the retrieval error is about 0.01-0.02. Therefore, it can 123 provide aerosol characteristic parameters with high accuracy and verify aerosol parameters 124 retrieved by satellite (Lu et al., 2019). In this study, AERONET AOD values are used as truth 125 values to verify the accuracy of satellite remote sensing AOD retrieval. AERONET AOD data is 126 classified into three levels: Level 1.0 (unfiltered), level 1.5 (cloud screening and quality control) 127 and level 2.0 (quality assurance). The study area mainly covers the Yangtze River Delta, and there 128 are several ground observation sites in the area. However, only two observation sites, Taihu and

129	Auznou-COMT sites have continuous observation data, while the duration of other sites is
130	relatively limited. Therefore, AERONET Level 1.5 AOD data with a large number of observed
131	values were selected to verify the retrieval results of MISR AOD.

132 **3. Methodology**

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133 **3.1** Problems in the previous surface reflectance estimation method

134 Accurate estimation of surface reflectance is a crucial and challenging aspect in the retrieval 135 of AOD from satellite remote sensing data (Remer et al., 2009; Gupta et al., 2016). Previous research has identified the variation patterns of 9-angle MISR AOD (Chen et al., 2021). However, 136 137 the AOD retrieved from 9 angles shows relatively large errors compared to AERONET AOD 138 (Table S3). Atmospheric correction can eliminate the effects of clouds and aerosols and obtain true 139 surface reflectance. The reflectance of MISR is calculated by inputting geometric parameters, 140 AOD, aerosol type, sensor radiation brightness data and sensor height into 6S model. In this study, 141 we input MISR geometric parameters and radiation data corresponding to Taihu and 142 Xuzhou-CUMT sites, and AOD parameters are AERONET AOD values of these two sites. The 143 reflectance of atmospheric correction for each pixel is calculated by the linear atmospheric 144 correction formula. To explore the reasons for the high AOD value of the retrieval from 9 MISR 145 angles, the MISR atmospheric correction reflectance and MISR surface reflectance were compared at the corresponding locations of the two sites (Figure 1) (MISR surface reflectance 146 147 calculation method referenced Chen et al. (2021)). The results show that the surface reflectance of 148 MISR is lower than the atmospheric reflectance of MISR. The MISR AOD retrieved is higher 149 compared to the AERONET AOD. Therefore, it is necessary to establish a correction model to



150 improve the MISR surface reflectance and improve the retrieval accuracy of MISR AOD.

Figure 1. Comparison of MISR surface reflectance and atmospheric corrected reflectance in blue
band (At the pixel locations of Taihu and Xuzhou-CUMT sites).

153 **3.2 Improved surface reflectance estimation method**

- 154 To construct an improved surface reflectivity correction model, the specific steps are shown
- as follows:
- 156 (1) Calculation of MODIS atmospheric corrected reflectance

157 The MODIS L1B data is corrected by the 6S model, and the MODIS atmospheric correction

158 reflectance is obtained. When using 6S model to calculate MODIS atmospheric correction

- 159 reflectivity, it is necessary to provide geometric parameters, AOD, atmospheric mode, aerosol type,
- 160 sensor radiance data, sensor height, spectral parameters and other parameters. In this study,
- 161 MODIS geometric parameters and radiance data corresponding to Taihu and Xuzhou-CUMT sites

162 are used. The AOD parameters were obtained from AERONET AOD data at both sites.

- 163 Atmospheric models select mid-latitude winter and mid-latitude summer to account for seasonal
- 164 variations in atmospheric transport. The aerosol type selected in this study is continental aerosol.
- 165 The sensor height is set to the observation height of the satellite. The spectral parameters are

166 defined according to the band of the MODIS sensor. By providing these parameters, we can

- 167 calculate the atmospheric correction reflectance of the MODIS sensor using the 6S model.
- 168 (2) Surface bidirectional reflectance Calculation
- 169 MODIS BRDF/Albedo product MCD43A1 data and Ross-Li model were used to simulate

surface bidirectional reflectance under MODIS and MISR observation geometry. The linear kernel
driven BRDF model includes three basic parameters: the reflectance of the lowest point view and
the weighting coefficient of the two kernel functions. The model can be calculated using formulas

173 1-9.

174
$$BRDF(\theta_s, \theta_v, \phi) = f_{iso}(\Lambda) + f_{vol}(\Lambda)K_{vol}(\theta_s, \theta_v, \phi) + f_{geo}(\Lambda)K_{geo}(\theta_s, \theta_v, \phi)$$
(1)

175
$$K_{vol}(\theta_s, \theta_v, \phi) = \frac{(\pi/2 - \xi)\cos\xi + \sin\xi}{\cos\theta_s + \cos\theta_v} - \frac{\pi}{4}$$
(2)

176
$$K_{geo}(\theta_s, \theta_v, \phi) = O(\theta_s, \theta_v, \phi) - \sec \theta'_s - \sec \theta'_v + \frac{1}{2}(1 + \cos \xi') \sec \theta'_s \sec \theta'_v$$
(3)

177
$$O(\theta_s, \theta_v, \phi) = \frac{1}{\pi} (t - \sin t \cos t) (\sec \theta'_s + \sec \theta'_v)$$
(4)

178
$$\cos t = \frac{h}{b} \frac{\sqrt{D^2 + (\tan \theta'_s \tan \theta'_v \sin \phi)^2}}{\sec \theta'_s + \sec \theta'_v}$$
(5)

179
$$D = \sqrt{\tan^2 \theta'_s + \tan^2 \theta'_v - 2\tan \theta'_s \tan \theta'_v \cos \phi}$$
(6)

180
$$\cos \xi' = \cos \theta'_s \cos \theta'_v + \sin \theta'_s \sin \theta'_v \cos \phi$$
(7)

181
$$\theta'_{s} = \tan^{-1}(\frac{b}{r}\tan\theta_{s})$$
(8)

182
$$\theta'_{\nu} = \tan^{-1}(\frac{b}{r}\tan\theta_{\nu})$$
(9)

In the above equations,
$$BRDF(\theta_s, \theta_v, \phi)$$
 is the surface bidirectional reflectance, while θ_s ,
 θ_v , and ϕ are the solar zenith angle, the satellite zenith angle, and relative azimuth angle,
respectively. A is bandwidth, while $K_{vol}(\theta_s, \theta_v, \phi)$ and $K_{geo}(\theta_s, \theta_v, \phi)$ are volumetric scattering
kernel and geometric optical scattering kernel, respectively, which are functions of the angle of

187 incidence and the angle of observation. f_{iso} , f_{vol} , and f_{geo} are the weights assigned to 188 isotropic scattering, volumetric scattering, and geometric optical scattering, respectively. ξ is the 189 scattering angle, while b, h, and r represent vertical radius of the sphere, horizontal radius of the 190 sphere and height of the center of the sphere, respectively. These three parameters can be defined 191 as fixed values. In the production process of the MODIS BRDF model, the relationship between h, 192 b and r parameters is h/b=2 and b/r=1 (Schaaf et al., 1999). According to the above equation, the 193 surface bidirectional reflection under any incident direction of the sun and observation direction of 194 the satellite can be obtained by extrapolating the nucleus. 195 (3) Improved MISR surface reflectance calculation

The MODIS atmospheric correction reflectance is brought into Eq. 10 and Eq. 11, and a new MISR surface reflectance estimate based on MODIS atmospheric correction is calculated. Regression fitting of MISR surface reflectance with the newly estimated MISR surface reflectance was performed (60% of the total sample data was randomly selected), and the surface reflectance error correction model was finally established, as shown in the following equation:

201

$$\rho(\theta_s, \theta_v, \phi)_{MISR_a} = \rho(\theta_s, \theta_v, \phi)_{MODIS_at} \times \frac{BRDF(\theta_s, \theta_v, \phi)_{MISR}}{BRDF(\theta_s, \theta_v, \phi)_{MODIS}}$$
(10)

In Eq. 10, $BRDF(\theta_s, \theta_v, \phi)_{MISR}$ and $BRDF(\theta_s, \theta_v, \phi)_{MODIS}$ are BRDFs obtained under MISR and MODIS angles, respectively. θ_s is the solar zenith angle, θ_v is the satellite zenith angle, and ϕ is the relative azimuth angle. $\rho(\theta_s, \theta_v, \phi)_{MISR_a}$ is the surface reflectance of MODIS at the geometric observation angle of MISR, and $\rho(\theta_s, \theta_v, \phi)_{MODIS_at}$ is the MODIS atmospheric corrected reflectance.

207 In this study, spectral data containing the typical characteristics of 28 different types of

vegetation, soil and water bodies were selected from 5 standard spectral libraries included with
ENVI software. The surface reflectance of MODIS and MISR sensors in the blue band is
calculated by the formula (Chen et al., 2021).

$$\rho(\theta_s, \theta_v, \phi)_{MISR} = \rho(\theta_s, \theta_v, \phi)_{MISR \ a} \times 0.9834 - 0.0081 \tag{11}$$

212 The New MODIS surface reflectance ($\rho(\theta_s, \theta_v, \phi)_{MISR_a}$) at the MISR angle obtained from

Eq. 10 is converted to the MISR surface reflectance by Eq. 11.

214 The MISR surface reflectance estimated by Eq. 11 was transformed using an error correction 215 model to obtain an improved MISR surface reflectance. The improved MISR surface reflectance 216 will be used for AOD retrieval. The MISR surface reflectance previously estimated based on 217 MODIS V5.2 algorithm was fitted with the MISR surface reflectance estimated based on MODIS atmospheric correction (60% of the data were randomly selected) by linear regression, and a 218 219 MISR correction model was established, as shown in Eq. 12. The previously estimated 9-angle 220 MISR surface reflectance is corrected by Eq. 12, and the improved surface reflectance of the 221 MISR sensor at 9 angles is obtained, and finally used for MISR AOD retrieval at 9 angles.

222

$$\rho(\theta_s, \theta_v, \phi)_{MISR-b}^{*} = \rho(\theta_s, \theta_v, \phi)_{MISR} \times 0.9209 + 0.0409$$
(12)

223 Where ρ_{MISR-b}^{*} is the improved MISR surface reflectance in Eq. 12.

224 **3.3 Flow of improved multi-angle AOD retrieval**

225	The flow of the improved surface reflectance algorithm in this study is shown in Figure 2.
226	Firstly, the atmospheric correction of MODIS L1B data is performed using 6S model. Then, the
227	MISR surface reflectance estimated by previous study was combined with the new MISR surface
228	reflectance estimated by Eq. 11 to build a MISR error correction model for obtaining the improved

229 MISR surface reflectance (Chen et al., 2021). The study retrieved the MISR AOD for nine camera 230 angles using improved MISR surface reflectance. We validated the improved MISR AOD with 231 AERONET AOD. We compared the improved AOD with the previously retrieved AOD and analyzed the accuracy and spatial distribution trends of the improved AOD. The AOD retrieval 232 233 method used in this study is based on chen et al. (2021). In our study, we used continental aerosols 234 for AOD retrieval and 6S model for atmospheric correction. Choosing the right aerosol type is crucial for obtaining accurate aerosol optical depth. Previous studies have shown that continental 235 236 aerosols can be used to estimate aerosol optical depth in the Yangtze River Delta (He et al., 2015). 237 In this study, the same aerosol type was used for AOD retrieval and atmospheric correction, and

the error transfer caused by aerosol type and atmospheric correction was not considered.





Figure 2. Flow chart of the improved MISR surface reflectance algorithm

241 **4. Results and discussion**

242 4.1 Improved MISR surface reflectance variation characteristics

The variation characteristics of MISR surface reflectance, MISR atmospheric correction 243 244 reflectance and improved MISR surface reflectance are shown in Figure 3. These values are the 245 average values of all sample data collected at the corresponding locations of the two sites in Taihu 246 and Xuzhou-CUMT during the period of validity from 2016-2018. At Taihu and Xuzhou-CUMT sites, the improved MISR surface reflectance of 9 camera angles is generally higher than that of 247 248 MISR surface reflectance, and lower than that of MISR atmospheric correction reflectance. The 249 surface reflectance of 9 camera angles MISR is between 0.02 and 0.04. On average, the improved 250 surface reflectance is higher than the previously estimated MISR surface reflectance.



Figure 3. Comparison of MISR surface reflectance, atmospheric corrected reflectance and improved surface reflectance in the blue band (This is multi-year averaged sample data from two sites, Taihu and Xuzhou-CUMT).

254

To clarify the trend of the improved surface reflectance, the study conducted a time-series analysis of the MISR surface reflectance and the improved surface reflectance. The improved

257 MISR surface reflectance is always higher than the previously estimated MISR surface reflectance

258 (Figure 4). The MISR surface reflectance is generally between 0 and 0.05, while the improved

- surface reflectance values is about 0.05 to 0.1. Overall, the improved surface reflectance was
- 260 increased.





Figure 4. Time series of surface reflectance in the blue band of the MISR sensor for 9 observation
angles (Table S2 lists the time series of the 9 angles from front to back)

263 4.2 Improved MISR AOD retrieval results

264 Three years of MISR AOD results were obtained using improved surface reflectance retrieval. 265 This study gives the results of the retrieval of the MISR sensor on June 12, 2018 for nine camera 266 observation angles (Figure 5). From the spatial distribution of AOD, it can be seen that the 267 retrieval results in the study area do not exceed 1. The overall spatial distribution trend is 268 generally consistent with the pre-improvement results (Chen et al., 2021), but there are differences 269 in the magnitude of the values. The values in the northeast and south range from 0.5 to 1, 270 indicating poor air quality in these areas. The AOD values retrieved from 5 camera observation 271 angles Ba, Aa, An, Af, and Bf are in the approximate range of 0.25-0.5. In the central region, most

of the AOD values from the 4 camera observation angles Da, Ca, Cf and Df are in the range of 0-0.25. The data show that the air quality in the central region is generally good, and the air pollution level is light in some areas. The higher AOD in southern Shandong and northern Jiangsu may be related to the increased of local aerosol emissions caused by human activities.



Figure 5. June 12, 2018 Improved MISR AOD retrieval results at 550 nm

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The improved MISR AOD spatial distribution results were then validated by comparing them with the MODIS AOD product. The MODIS AOD product has a resolution of 3km. The trends in the spatial distribution of MODIS AOD products are consistent with the improved MISR AOD (Figure 6). However, the MODIS AOD product has more missing data, and the AOD obtained by using the improved algorithm for retrieval can avoid missing data. In addition, the AOD retrieval with the improved algorithm has a higher resolution compared to the image quality of the MISR AOD product.



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Figure 6. MODIS AOD 550nm product spatial distribution on June 12, 2018

285 4.3 Verification of improved MISR AOD

There are many AERONET sites in the Yangtze River Delta region, but currently only two sites, Taihu and Xuzhou-CUMT, are able to provide continuous data, and the other sites acquire data on limited time scales. Therefore, we selected the Taihu and Xuzhou-CUMT sites with more data for verification. To validate the retrieved MISR AODs, we selected valid AOD records in the 550nm band within a 30-minute interval between the AERONET ground observation site and the Terra satellite. The MISR sensor 9 camera views take about 7 minutes to be able to observe the same geographic location at relatively short intervals. Therefore, we used the AERONET AOD averages as approximate truth values and compared them with the retrieved MISR AODs to validate and minimize errors due to time lag. Spatially, we selected the image elements observed by the MISR sensor from nine angles and compared them with the corresponding positional AERONET observations. Since the sun photometer did not have a wavelength of 550nm corresponding to the retrieval results, the AOD at 550nm was calculated by applying Angstrom (Eq. 13).

299

$$\tau(\lambda) = \beta \lambda^{-\alpha} \tag{13}$$

300 In the formula, $\tau(\lambda)$ is the AOD at wavelength λ , β is the concentration of aerosols 301 throughout the atmosphere, and α is the wavelength index of Angstrom.

302 In this study, four parameters, correlation coefficient (R), root mean square error (RMSE), 303 p-value and relative mean bias (RMB), will be used to assess the accuracy of the remotely sensed 304 AOD dataset. The specific calculation principles for the three parameters R, RMSE and RMB are 305 shown in Eq. 14-16. The validation results of AOD dataset of Taihu and Xuzhou-CUMT sites 306 during 2016-2018 after improvement in this study are shown in Figure 7 and Figure 8. 307 Generally, the scatter plots are distributed above and below the 1:1 line. R is a parameter 308 characterising the correlation between the remote sensing retrieval results and the ground retrieval 309 results. R reached 0.89 for the Taihu site and 0.85 for the Xuzhou-CUMT site. RMSE is a 310 parameter characterising the absolute error of the remote sensing retrieval results, and the 311 minimum RMSE is 0.21 for the Taihu site. The minimum RMSE is 0.20 for the Xuzhou-CUMT 312 site. RMB is a parameter used to describe the relative error of remote sensing retrieval results, and

the minimum RMB is 0.52 for the Taihu site and 0.32 for the Xuzhou-CUMT site. Overall, the

314 retrieval results at angle An are optimal after algorithm improvements (Taihu: R=0.84, RMB=0.52; 315 Xuzhou-CUMT: R=0.85, RMB=0.47). By comparing the results with the pre-improvement AOD 316 validation, the improved AOD retrieval accuracy has been significantly improved (Table 1). In 317 Table 1, R is the correlation between the AOD retrieval results and AERONET AOD before 318 improvement. RMB is the relative deviation between the AOD retrieval by the pre-improvement 319 algorithm and AERONET AOD. Improved R is the correlation between the improved AOD 320 retrieval results and AERONET AOD. Improved RMB is the relative deviation between the AOD 321 retrieved with the improved algorithm and AERONET AOD.

322
$$R = \frac{\sum_{i=1}^{N} (A_i - \overline{A}) (A_i' - \overline{A'})}{\sqrt{\sum_{i=1}^{N} (A_i - \overline{A})^2 \sum_{i=1}^{N} (A_i' - \overline{A'})^2}}$$
(14)

323
$$RMSE = \sqrt{\sum_{i=1}^{N} (A_i - A_i)^2 / N}$$
(15)

324
$$RMB = \sum_{i=1}^{N} (A_i - A_i) / N$$
(16)

325 where A_i is the retrieve MISR AOD, A'_i is the corresponding AERONET AOD, \overline{A}

and A' are the mean values of the retrieve MISR AOD and AERONET AOD, respectively. N is

327 the number of valid matching results for AERONET AOD and MISR AOD.





328 Figure 7. Comparison between improved MISR AOD and AERONET AOD at Taihu site (N is

329

the number of verification points, red line represents a linear fitting line)





330 Figure 8. Comparison between improved MISR AOD and AERONET AOD at Xuzhou-CUMT

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site (N is the number of verification points, red line represents a linear fitting line)



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Figure 9. Comparison of validation of retrieved AOD with MODIS AOD product

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and after improvement

Table 1. Precision comparison of MISR AOD and AERONET AOD before

Site	Angle	R	RMB	Improved R	Improved RMB
	Da	0.77	1.08	0.80	0.69
	Ca	0.70	1.00	0.72	0.74
	Ba	0.77	0.61	0.80	0.70
	Aa	0.81	0.68	0.82	0.71
Taihu	An	0.45	1.22	0.84	0.52
	Af	0.72	0.87	0.89	0.68
	Bf	0.72	0.60	0.82	0.67
	Cf	0.57	0.65	0.79	0.73
	Df	0.77	0.47	0.79	0.66
Xuzhou-	Da	0.45	1.58	0.65	0.61

CUMT	Ca	0.59	0.96	0.71	0.82
	Ba	0.67	0.78	0.70	0.32
	Aa	0.73	0.78	0.82	0.49
	An	0.75	0.85	0.85	0.47
	Af	0.72	0.63	0.74	0.62
	Bf	0.62	0.65	0.75	0.53
	Cf	0.68	0.66	0.72	0.54
	Df	0.67	0.65	0.72	0.53

³³⁶

337 Comparing the validation results of the MODIS AOD product with those of the observation sites (Taihu: R=0.59, RMSE=0.19, P<0.05, RMB=0.52; Xuzhou-CUMT: R=0.71, RMSE=0.25, 338 P<0.05, RMB=0.44) (Chen et al., 2021), we found that the Taihu and Xuzhou-CUMT sites have 339 340 high correlation between the improved MISR AOD and MODIS AOD products. The smaller 341 observation angle of the improved MISR AOD, the closer the error is to the MODIS AOD product. 342 The observation angle of MISR An is the same as that of MODIS. Therefore, we selected two 343 sites for validation at the An observation angle and MODIS AOD product (Figure 9). The results 344 show that the An AOD retrieval by the improved algorithm correlates well with the MODIS AOD 345 product. The position errors of the two image elements are close to each other. The RMSE of the 346 Xuzhou-CUMT site is slightly higher than that of Taihu site, and the RMB of Taihu site is slightly higher than that of the Xuzhou-CUMT site. 347

348 5. Conclusion

In this study, we first explored the problems of estimating surface reflectance in our previous study. We obtained an error correction model for surface reflectance by using a linear fit of the MISR surface reflectance and a new estimate of the MISR surface reflectance. The improved MISR surface reflectance was obtained through the error correction model. We then retrieved a new AOD product using the improved surface reflectance and a lookup table constructed from the
6S model. Two AERONET ground observation sites with longer time series were used to validate
the AOD obtained by satellites.

(1) Overall, the improved AOD and its spatial distribution trends are consistent with our
previous results. The AOD estimated by improved method exhibited higher accuracy and a high
degree of agreement with the AERONET ground-based observational AOD. Overall, the retrieval
results at angle An are optimal after algorithm improvements (Taihu: R=0.84, RMB=0.52;
Xuzhou-CUMT: R=0.85, RMB=0.47). The improvements at the Xuzhou-CUMT site are more
pronounced compared to the Taihu site.

362 (2) More importantly, the optimal AOD obtained in this study has fewer missing pixels and
363 finer spatial resolution than the MODIS AOD product. Through the validation with MODIS AOD
364 products, the AOD of An optimal angle obtained by the improved algorithm is highly correlated
365 with MODIS AOD products.

366 (3) In future, more aerosol models that conform to the actual situation in the study area can 367 be constructed using the AERONET ground observation data and introduced into the MISR AOD 368 retrieval algorithm to further improve the accuracy of the AOD retrieval results. In this study, the 369 AERONET AOD was used as the true value and as an input for the AOD parameter in the 6S model for atmospheric correction of MISR and MODIS images. We then obtained a surface 370 371 reflectivity error correction model to retrieve AOD for the entire region. It should be emphasized that the more AERONET sites used to train the corrected model, the more accurate AOD retrieved 372 373 may be obtained by this method. However, the data from the AERONET ground observation sites 374 were limited. In future, the study area can be expanded on a large scale and for a longer time 375 series.

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381 Competing interests

382 The contact author has declared that none of the authors has any competing interests.

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