

Supplementary Information for

MODIS Collection 6.1 aerosol optical depth products over land and ocean: validation and comparison

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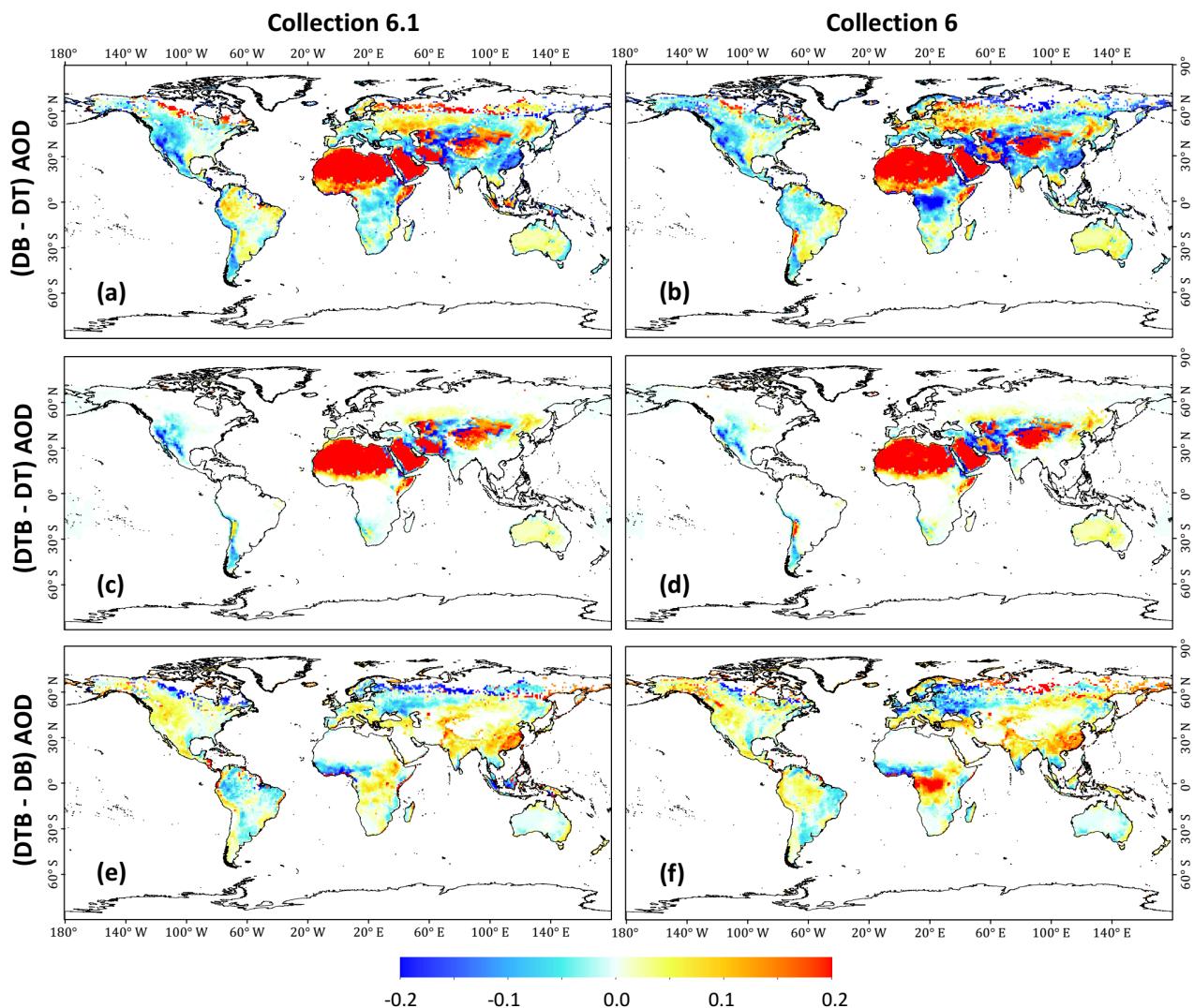


Figure S1. Mean differences in C6.1 (left panels) and C6 (right panels) annual mean AODs between (a, b) DB and DT, (c, d) DTB and DT, and (e, f) DTB and DB. Data are from 2013–2017.

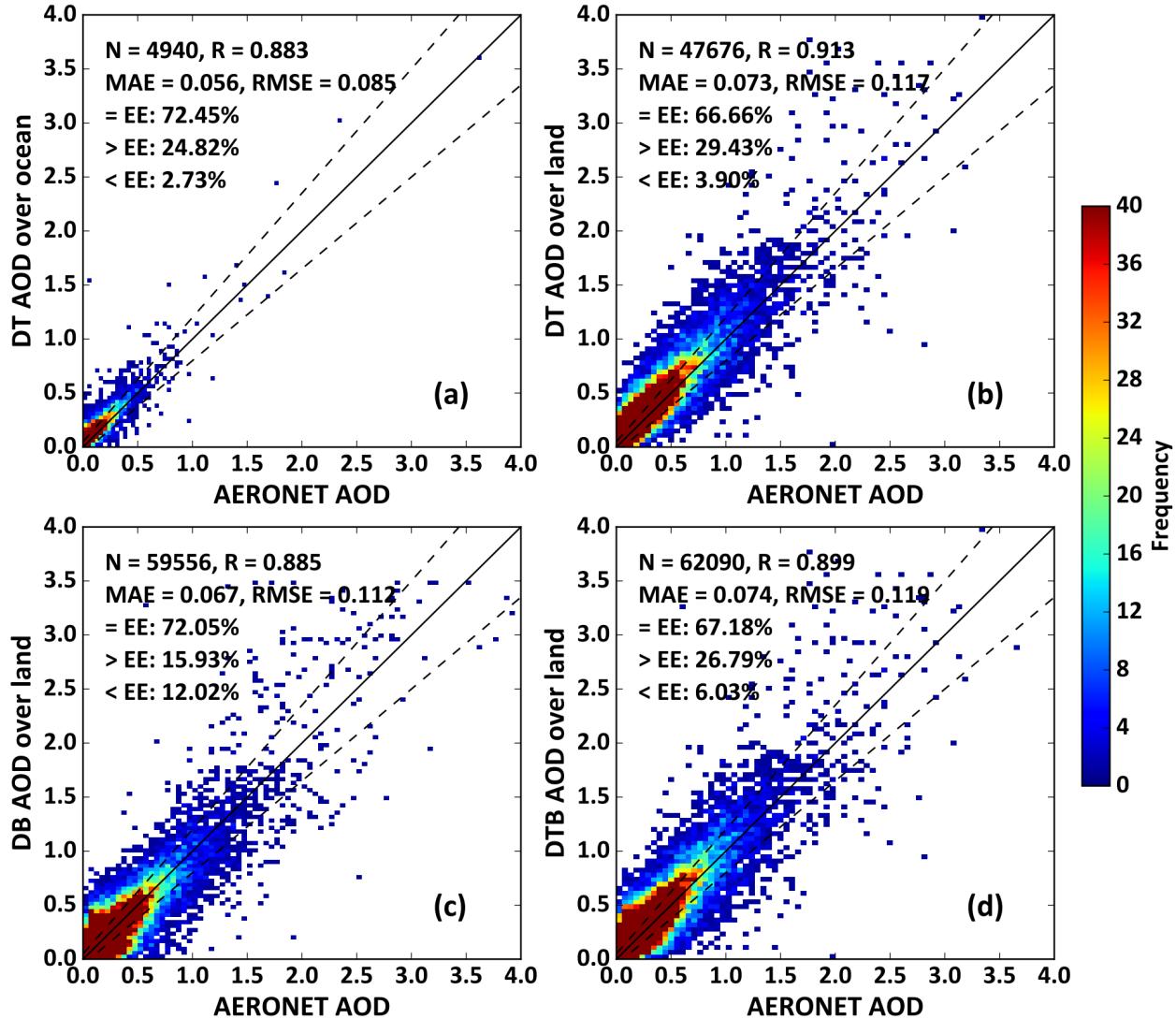


Figure S2. Density scatter plots of Terra MODIS C6 DT over ocean (a), DT (b), DB (c) and DTB (c) AOD retrievals over land against AERONET AOD measurements from 2013 to 2017. The solid line denotes the 1:1 line, and the dashed lines denote the envelope of the expected error (EE). The sample size (N), correlation coefficient (R), mean absolute error (MAE), and root-mean-square error (RMSE) are also given.

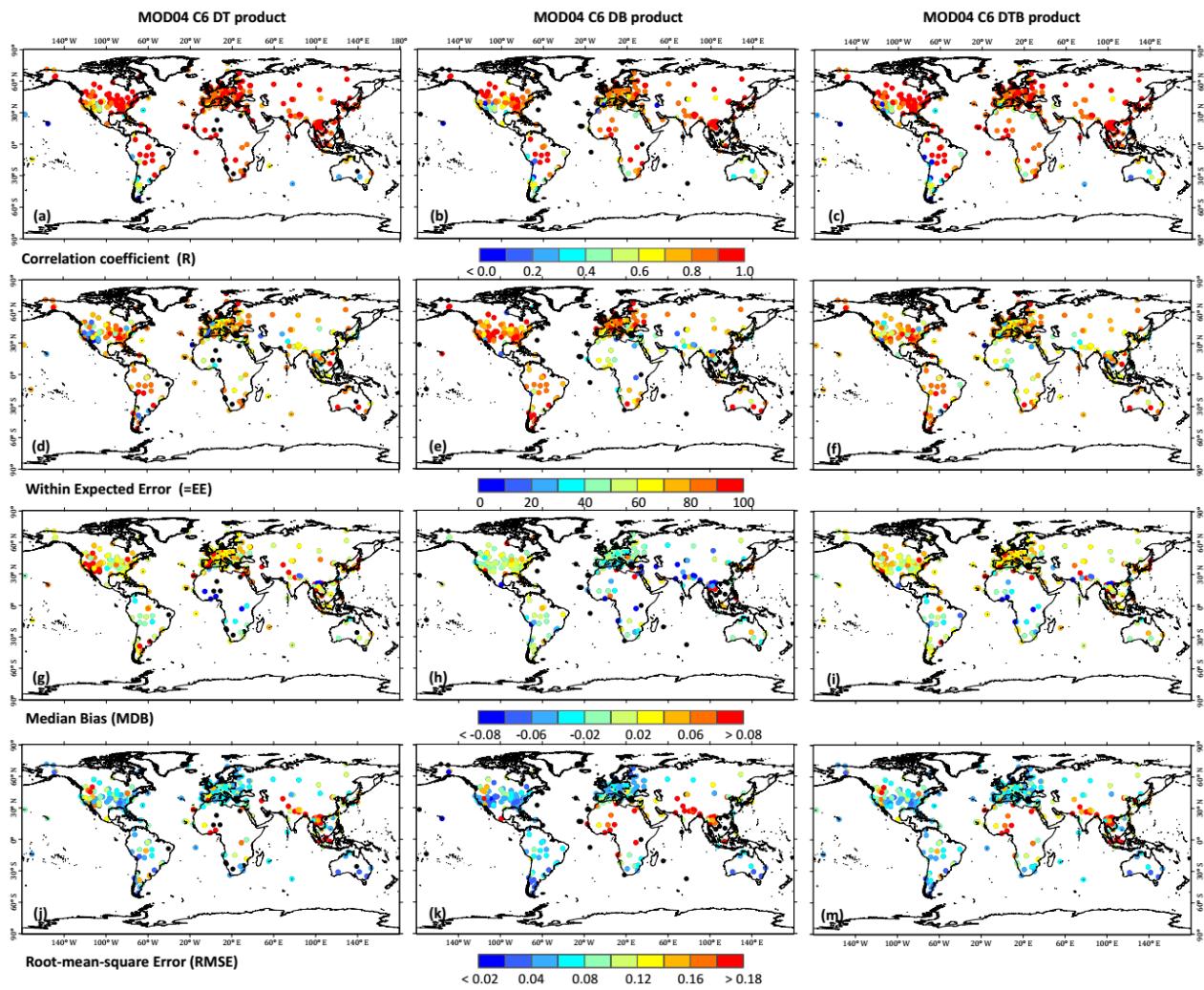


Figure S3. Validation of Terra MODIS C6 DT, DB, and DTB AOD retrievals against AERONET AODs for each site from 2003 to 2017: (a-c) correlation (R), (d-f) percentage of retrievals within the expected error envelope (%), (g-i) median bias, and (j-m) root-mean-square error.

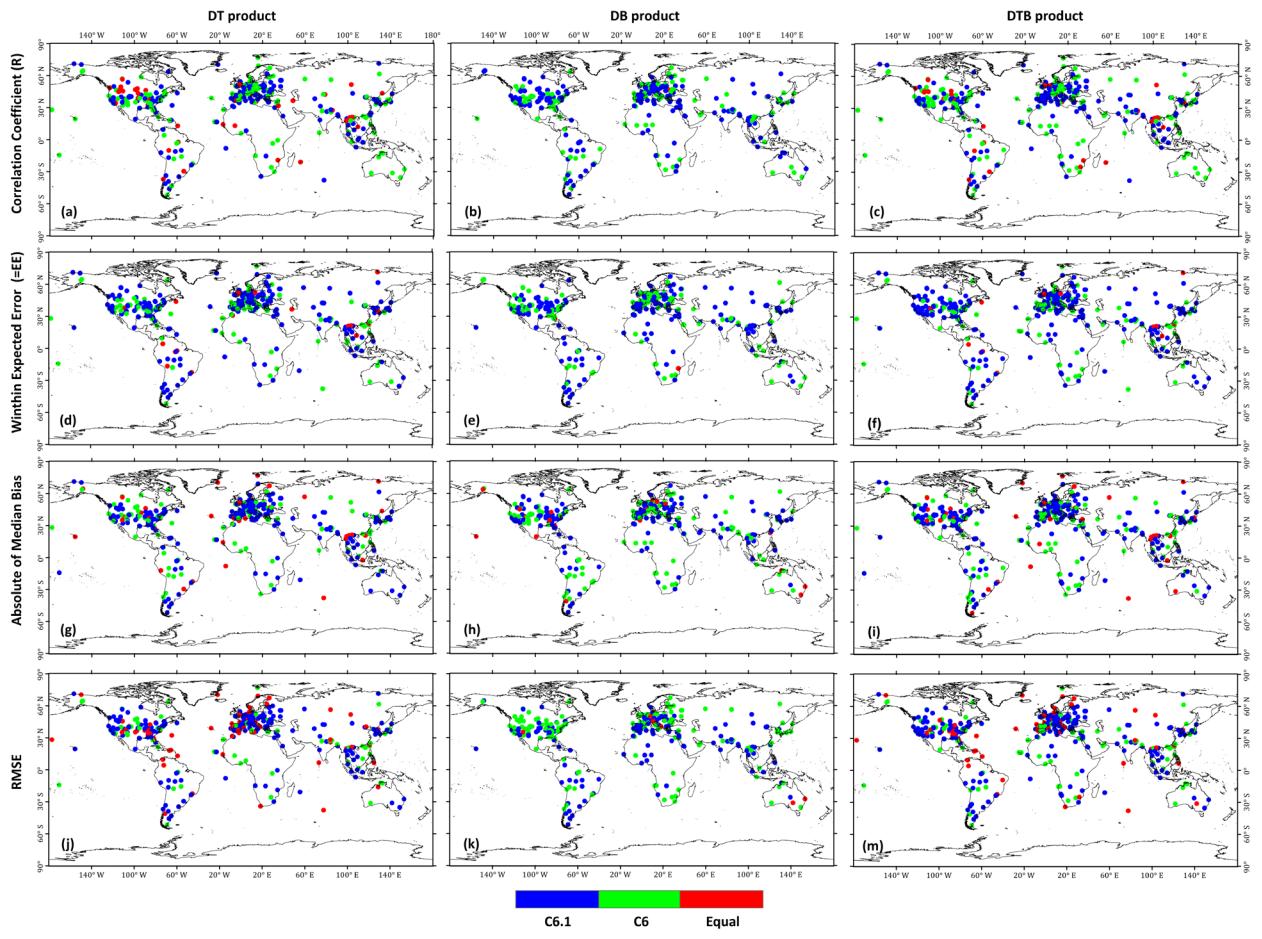


Figure S4. Maps showing the best performance of MODIS DT, DB and DTB algorithms across each site between C6.1 and C6 by different statistical metrics: (a) data collections, (b) percentage of collections falling within the EE, (c) median bias and (d) RMSE

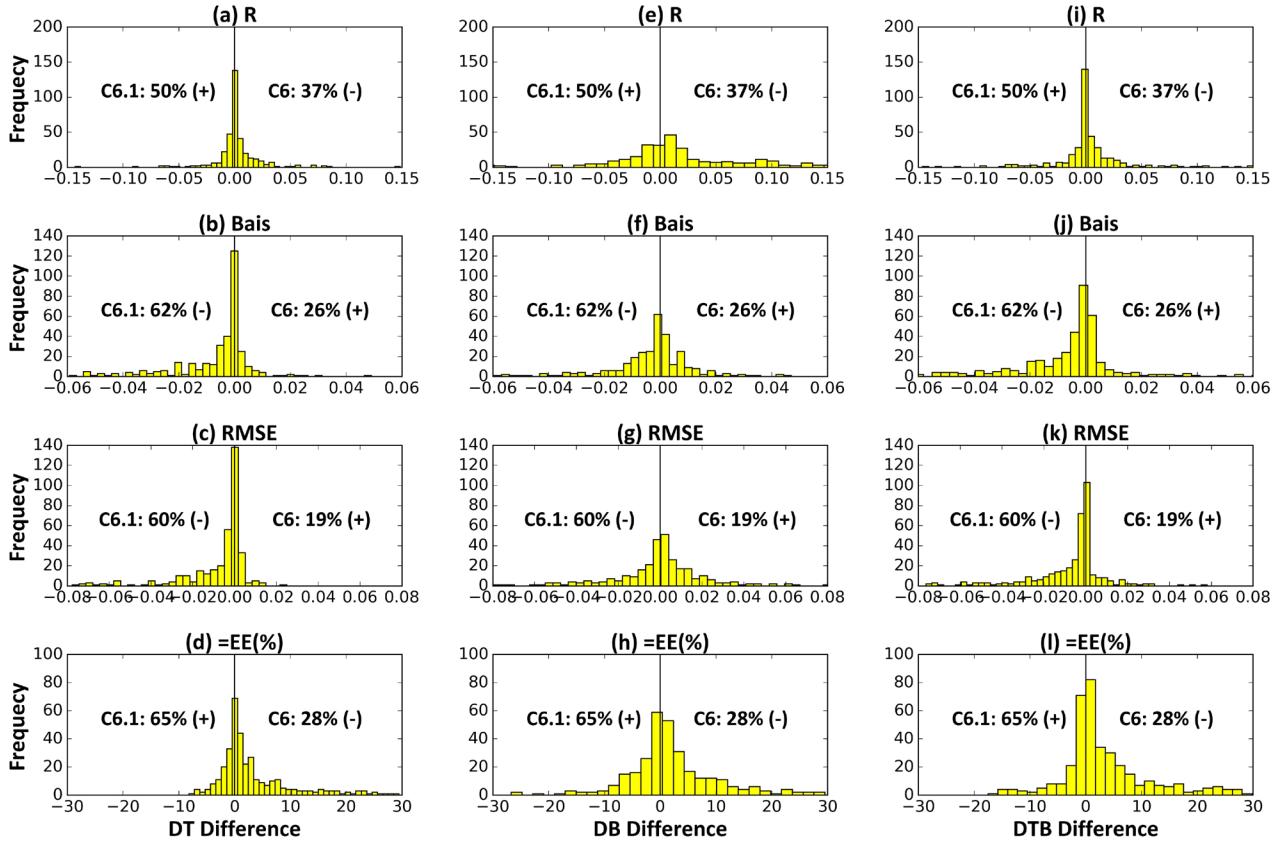


Figure S5. Frequency histograms of (C6.1 DT – C6 DT, a-d), (C6.1 DB – C6 DB, e-h) and (C6.1 DTB – C6 DTB, i-l) differences in validation statistics for the ensemble of sites with at least 20 retrievals. The statistics are the correlation coefficient (R), the absolute value of the median bias, the root-mean-square error (RMSE), and the percentage of retrievals within the EE envelope (=EE, %). For each metric, the fraction of sites where each algorithm performs best is shown in each panel. The annotation represents the proportions for validation statistics, where the plus and minus signs in parentheses indicate the negative and positive values.

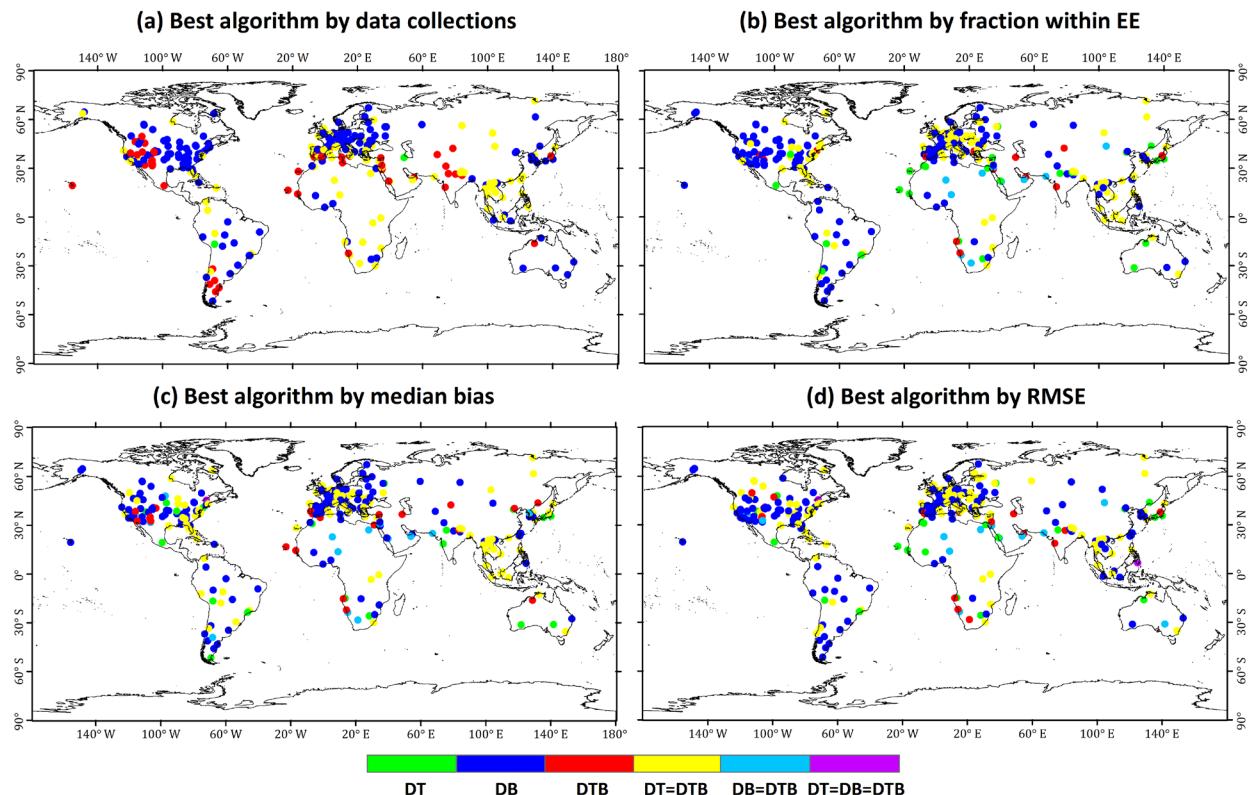


Figure S6. Geographical distribution of sites showing where the MODIS C6 DT, DB, and DTB algorithms performed the best on land according to different statistical metrics: (a) the number of retrievals, (b) the percentage of points falling within the EE envelope, (c) the median bias, and (d) the root-mean-square error (RMSE).

Table S1. Review of related studies on aerosol retrieval algorithms for satellite sensors

Satelite	Full name	Related study
AATSR	Advanced AlongTrack Scanning Radiometer	Veefkind & de Leeuw, 1998; North et al., 1999
AVHRR	Advanced Very High Resolution Radiometer	Li et al., 2013; Riffler et al., 2010
Gaofen	-	She et al., 2017; Sun et al., 2017
GOCI	Geostationary Ocean Color Imager	Choi et al., 2016
HJ-1	Huanjing-1	Sun et al., 2010; Li et al., 2012
Himawari-8	-	Ge et al., 2018; Yan et al., 2018
Landsat	-	Lyapustin, et al., 2003; Wei et al., 2017
MERIS	Medium-resolution imaging spectrometer	Santer et al., 2007; Ramon & Santer, 2001
MISR	Multi-angle Imaging Spectro-Radiometer	Kahn et al., 2010; Martonchik et al., 1997
MODIS	Moderate Resolution Imaging Spectro-radiometer	Kaufman et al., 1997; Levy et al., 2007; Hsu et al., 2004, 2006.
OMI	Ozone Monitoring Instrument	Livingston et al., 2009; Torres et al., 2007
SeaWiFS	Sea-Viewing Wide Field-of-View Sensor	Sayer et al., 2012
TOMS	Total Ozone Mapping Spectrometer	Hsu et al., 1999; Torres et al., 2002
VIIRS	Visible Infrared Imaging Radiometer Suite	Jackson et al., 2013; Liu et al., 2014

Table S2. Statistics of annual mean aerosols for land, ocean and each region calculated from Terra C6.1 and C6 DT, DB and DTB AOD datasets from 2013 to 2017

Algorithm	DT		DB		DTB	
Collection	C6.1	C6	C6.1	C6	C6.1	C6
Ocean	0.162±0.070	0.159±0.070	-	-	0.162±0.070	0.159±0.070
Land	0.220±0.152	0.217±0.148	0.210±0.140	0.207±0.133	0.217±0.142	0.216±0.137
EAA	0.262±0.158	0.252±0.148	0.220±0.130	0.218±0.132	0.239±0.450	0.240±0.136
SAA	0.500±0.184	0.497±0.188	0.356±0.185	0.365±0.181	0.393±0.201	0.404±0.198
SEA	0.310±0.090	0.310±0.091	0.360±0.130	0.312±0.109	0.310±0.090	0.311±0.091
EUR	0.182±0.155	0.177±0.143	0.180±0.051	0.197±0.087	0.180±0.100	0.171±0.098
ENAM	0.157±0.037	0.159±0.037	0.155±0.041	0.157±0.045	0.158±0.037	0.161±0.037
WNAM	0.153±0.042	0.153±0.040	0.107±0.053	0.116±0.054	0.132±0.045	0.140±0.041
SAM	0.162±0.079	0.164±0.080	0.161±0.093	0.153±0.064	0.153±0.079	0.159±0.077
NAME	0.336±0.167	0.337±0.175	0.343±0.125	0.331±0.132	0.340±0.109	0.335±0.117
SAF	0.218±0.160	0.217±0.159	0.187±0.152	0.174±0.103	0.206±0.163	0.212±0.160
OCE	0.093±0.053	0.094±0.053	0.050±0.046	0.061±0.040	0.061±0.051	0.066±0.050

Table S3. Statistics describing the relationships between Terra C6 DT, DB, and DTB AOD retrievals and AERONET AOD measurements for each region from 2013 to 2017.

Region	N			R			MDB			RMSE			Within EE (%)		
	DT	DB	DTB	DT	DB	DTB	DT	DB	DTB	DT	DB	DTB	DT	DB	DTB
EAA	4449	4816	5738	0.915	0.933	0.931	0.054	0.010	0.045	0.171	0.146	0.160	59.2	64.3	63.0
SAA	2136	2316	2440	0.863	0.867	0.860	0.035	-0.083	0.008	0.201	0.218	0.200	62.1	49.2	58.9
SEA	2246	1866	2247	0.904	0.871	0.904	0.034	0.014	0.034	0.201	0.196	0.201	59.4	49.8	59.3
EUR	15494	16143	16465	0.845	0.770	0.842	0.039	-0.009	0.036	0.086	0.070	0.081	67.4	78.7	68.7
ENAM	8500	9020	8772	0.916	0.729	0.901	0.025	0.015	0.024	0.071	0.072	0.073	77.1	78.4	76.2
WNAM	6511	9420	9638	0.808	0.837	0.801	0.049	0.007	0.030	0.122	0.078	0.103	55.2	80.6	67.4
SAM	3157	3829	3736	0.916	0.925	0.930	0.024	0.002	0.013	0.094	0.059	0.079	67.5	84.1	80.0
NAME	1841	7842	8208	0.905	0.802	0.802	0.008	0.023	0.029	0.144	0.175	0.171	64.6	49.4	51.3
SAF	1329	1800	2035	0.908	0.823	0.862	-0.009	-0.014	-0.004	0.090	0.094	0.095	75.8	69.6	70.5
OCE	1394	2365	2190	0.664	0.461	0.594	0.004	-0.011	0.002	0.049	0.053	0.051	83.5	82.3	82.6

Table S4. Percentage of sites where a given algorithm performed best on land for a given metric. Data are from 2013 to 2017. The metrics are number of retrievals (N), correlation coefficient (R), estimated error (EE), median bias (MDB), and root-mean-square error (RMSE).

Collection	Collection 6.1					Collection 6				
Best in	N	R	EE	MDB	RMSE	N	R	EE	MDB	RMSE
DT	00.64	16.35	08.01	06.73	07.69	00.32	18.33	06.75	05.14	07.07
DB	51.92	24.68	49.04	48.08	47.44	45.66	21.86	53.70	52.41	52.73
DTB	17.63	09.94	06.41	07.69	00.00	16.40	08.36	07.40	08.04	00.00
DTB=DT	29.81	46.15	33.01	32.69	33.65	37.62	47.27	29.26	29.26	27.33
DTB=DB	00.00	02.88	03.53	04.49	04.49	00.00	03.22	02.89	04.18	05.14

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