Supplement of

Vegetation effects redistribute dust globally

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Section S1. Evaluation of threshold velocity in this study against the observationally retrieved threshold by Pu et al. (2020)

For the evaluation of the threshold velocity (u_t) in this study, we utilized the observationally retrieved 10-m wind speed threshold ($V_{threshold}$) based on dust aerosol optical depth (DAOD) from Pu et al. (2020). Specifically, we employed the version calculated using a DAOD criterion—0.5 for arid regions and 0.05 for semi-arid regions—to identify active dust emission events (Pu et al., 2020). To ensure a consistent point-to-point comparison, threshold velocity (u_t) used in this study with an original spatial resolution of 2.5° longitude × 1.27° latitude, was regridded to the 0.5° × 0.5° spatial resolution of the observational dataset (Fig. S2a–c).

A comparison with satellite-retrieved thresholds ($V_{threshold}$) (Pu et al., 2020) showed that u_t has an overall alignment with $V_{threshold}$ in major desert regions, such as North Africa and Middle East, but exhibited much higher values in vegetated secondary sources, such as North America and southern Africa, especially when the vegetation effect is considered (Fig. S2d–e).

The discrepancy between the threshold velocity (u_t) used in this study and the satellite-retrieved ($V_{threshold}$) from Pu et al. (2020) primarily reflects fundamental differences in methodological definitions. In our scheme, u_t is determined from soil properties and a physically based parameterization of vegetation effects, representing the local conditions required for dust emission. In contrast, $V_{threshold}$ is inferred from dust aerosol optical depth (DAOD) associated with dust events. Because DAOD integrates contributions from both local emissions and dust transported from distant sources, it may not exclusively reflect local emission processes. The diagnostic utility of DAOD in representing local emission sources remains a subject of ongoing discussion. For example, some recent studies (Chappell et al., 2023) show limited correlation of DAOD with observed dust point sources (DPS), whereas others (Mahowald et al., 2024) argue that such discrepancies may result from the spatial and temporal scales of the satellite observations rather than a lack of physical connection. Consequently, the divergence between $V_{threshold}$ and adjusted u_t in this study underscores the inherent challenge of reconciling local-scale surface dynamics with column-integrated atmospheric constraints.

Section S2. Discussion of model–observation discrepancies in DAOD

In this study, dust-climate interactions were decoupled (Sect. 2.3.2), thereby hindering the radiative feedback typically provided by dust aerosols. The atmospheric heating effect derived from dust absorption is a fundamental driver for the northward progression of the African Summer Monsoon (Balkanski et al., 2021). The resulting suppression of monsoon activity would reduce surface wind speeds, directly curtailing dust emissions in source regions like Mali/Niger and Bodele/Sudan during summer (JJA). This dynamic deficiency also limits the further transport or deposition (Pope et al., 2016) of dust to the African west coast and mid-Atlantic, leading to the underestimation in these regions. The underestimation at the African West Coast and the Mali/Niger border might also indicate the model has difficulties to resolve sharp land-ocean transitions and complex coastal boundary layer dynamics. Furthermore, large-scale models typically fail to resolve mesoscale convective systems (MCSs) and their associated cold pools (Bergametti et al., 2022). These systems trigger high-intensity dust events, such as haboobs (Pope et al., 2016), particularly during the pre-monsoon and early monsoon phases (May–June) when the soil is most vulnerable due to minimal vegetation cover. The inability to simulate these sub-grid convective gusts might also result in a negative bias in DAOD.

Similar to the patterns observed in North Africa, the simulated DAOD in the southern Middle East also exhibits underestimation during MAM and JJA, at the time part of this region is affected by the tropical monsoon (An et al., 2015). This discrepancy further reinforces the argument that monsoon-related dynamical deficiencies rooted in the decoupling of dust-climate interactions (Balkanski et al., 2021), and the poor parameterization of convection to capture haboob dust emission (Pope et al., 2016; Bergametti et al., 2022), might be important sources of model bias.

Unlike the tropical regions influenced by the summer monsoon, the decoupling of dust-climate interactions in arid inland regions likely suppresses the aerodynamic forces necessary for effective long-range transport. This may lead to an underestimation of the transport of Asian emission (Kok et al., 2021b), where emitted dust accumulates near the source region, thereby artificially magnifying the local DAOD, such as in the Kyzyl Kum desert during MAM and JJA, and Gobi during JJA (Fig. 6).

In contrast, the underestimation of DAOD in Taklamakan Desert (TD) especially during both MAM and DJF suggests a common model deficiency in resolving the vertical lofting of dust within this topographically enclosed basin (Nan and Wang, 2018). Surrounded by mountains exceeding 5 km, the TD acts as a trap where dust is primarily confined below the 4 km altitude threshold. This close-to-the-ground confinement is maximal during DJF, where only 1.9 % of dust reaches the 4–6 km layer, compared to 8.2 % in MAM (Nan and Wang, 2018). Our results indicate that during these cold and transition seasons, the model likely overestimates atmospheric stability or lacks the episodic convective energy required to lift dust across the 4 km boundary. In MAM, this translates into the inability of the model to bridge the “de-coupled” vertical layers (Nan and Wang, 2018), while in DJF, the stable stratification further suppresses dust mobilization and vertical mixing. Consequently, the inability to sustain dust particles above the threshold might lead to excessive near-source deposition and an overall depletion of the column-integrated DAOD.

Section S3. Model–Observation discrepancies in dust surface concentration

The downwind region of dust emission hotspots (e.g., sites over Japan in Fig. 2b, Fig. 9), the 4-mode configuration exhibits a more pronounced overestimation compared to the 1-mode. This discrepancy likely stems from the competing effects of massive initial emission and rapid gravitational settling of coarse particles. Although the 4-mode setup includes the 5–10 μm fractions that typically dominate observed spring dust events in Japan (Uematsu et al., 2003), the model may over-specify their initial mass flux. Because Japan is relatively close to Asian dust sources, the high surface concentrations might be driven by a massive initial emission flux that outpaces the removal rate over relatively short transport distances. While the 1-mode configuration's focus on finer dust fractions (MMD=2.5 μm) inherently omits dominant coarse particles, its closer agreement with observations appears physically fortuitous. Specifically, the slower settling velocity of the 2.5 μm dust fraction may artificially extend particle residence time to mask deficiencies in transport efficiency or an overestimation of emission. Ultimately, these systematic differences underscore that downwind dust concentrations are governed by the interplay between particle size representation and atmospheric lifetime, rather than emission strength alone.

Section S4. Analysis of persistent overestimation of dust deposition in Antarctica

Regarding the Southern Hemisphere, the persistent overestimation of dust surface total deposition over Antarctica—a common feature in global models (Leung et al., 2024)—likely reflects unresolved topographic and boundary-layer processes. Unlike the highly heterogeneous alpine terrain, where steep peaks can cause sub-grid elevation mismatches (Sect. 2.3.6), the Antarctic interior consists of a vast, relatively homogeneous plateau with a model-represented average elevation of approximately 2000 m (mean elevation south of 70° S is 1952 m), basically aligning with the description of a high-altitude plateau exceeding 2000 m (Tewari et al., 2021). The stations on this plateau remain within the shallow boundary layer resolved by the model and can thus be reasonably compared with simulated surface deposition. However, while the model captures the broad plateau elevation, it may not resolve fine-scale katabatic winds confined to this shallow, stably stratified layer, which regulate near-surface transport and dust divergence toward the coast. Coarse-resolution models cannot fully represent these flows, likely underestimating seaward export and enhancing inland deposition.

Table S1. Target and original emission from the vegetation-impact simulation in 1-mode configuration prior to rescaling at the year of 2008, along with the corresponding rescaling factors applied in the nine regions.

Regions	Target dust emission (Tg y^{-1})	Original dust emission (Tg y^{-1})	Rescaling factors
Western North Africa	314	448	0.7
Eastern North Africa	257	397	0.6
The Sahel	200	242	0.8
The Middle East*	210	79	2.7
Central Asia*	283	106	2.7
East Asia	207	365	0.6
North America	46	2	23.2
Australia	57	38	1.5
South America	68	1	82.0
Southern Africa	36	1	62.0

*Note: To provide a more detailed regional characterization, the Middle East and central Asia (originally aggregated in Kok et al. (2021a), Fig. S4) were differentiated: The Middle East is defined as the region situated south of 40°N and west of 60°E, and the remaining areas of the original domain, are categorized as central Asia.

Table S2. Global mean annual Dust Aerosol Optical Depth (DAOD) and statistical performance metrics for simulated DAOD, surface PM concentration, and dust deposition. Results are shown for both control and vegetation-impact simulations in the 1-mode and 4-mode configurations with different scaling coefficients for all data points. For 4-mode, the scaling coefficient refers to the Mode 2 and Mode 3, which is the same for both modes.

		1-mode		4-mode		1-mode		4-mode		4-mode		4-mode	
		Control	Vegetation-impact	Control	Vegetation-impact	Control	Vegetation-impact	Control	Vegetation-impact	Control	Vegetation-impact	Control	Vegetation-impact
Scaling coefficient		1		1		0.74		1.24		1.44		1.51	
Global mean annual DAOD		0.042	0.034	0.026	0.021	0.031	0.025	0.031	0.025	0.035	0.028	0.037	0.030
DAOD vs AERONET-SDA	R^2	0.68	0.63	0.68	0.68	0.68	0.63	0.67	0.68	0.67	0.68	0.67	0.68
	RMSE	0.16	0.13	0.07	0.05	0.09	0.07	0.10	0.07	0.13	0.09	0.14	0.09
	MB	0.13	0.09	0.03	0.00	0.06	0.03	0.06	0.03	0.09	0.05	0.10	0.06
	Slope	0.48	0.51	0.73	0.87	0.64	0.69	0.61	0.73	0.53	0.64	0.51	0.62
DAOD vs Kok et al. (2021a)	R^2	0.53	0.59	0.48	0.54	0.53	0.59	0.48	0.54	0.48	0.54	0.48	0.54
	RMSE	0.16	0.12	0.10	0.09	0.10	0.08	0.11	0.09	0.14	0.10	0.14	0.11
	MB	0.07	0.04	-0.01	-0.03	0.02	-0.01	0.02	0.00	0.04	0.02	0.05	0.02
	Slope	0.42	0.51	0.60	0.74	0.57	0.69	0.50	0.62	0.44	0.55	0.42	0.53
Dust surface concentration evaluation	R (log)	0.90	0.91	0.89	0.90	0.90	0.91	0.89	0.90	0.89	0.90	0.89	0.90
	RMSE (log)	0.45	0.43	0.50	0.51	0.44	0.45	0.51	0.50	0.53	0.51	0.54	0.51
	MB (log)	0.11	0.01	0.02	-0.09	-0.02	-0.12	0.11	0.00	0.17	0.06	0.19	0.08
Dust surface total deposition evaluation	R (log)	0.84	0.86	0.81	0.83	0.84	0.86	0.81	0.83	0.81	0.83	0.81	0.83
	RMSE (log)	0.77	0.70	0.86	0.80	0.74	0.69	0.90	0.82	0.93	0.84	0.94	0.85
	MB (log)	0.27	0.14	0.31	0.17	0.14	0.01	0.39	0.26	0.45	0.32	0.47	0.34

Table S3. Summary of statistics (R^2, R, RMSE, NRMSE, and MB) for the model evaluation, including dust emission, DAOD, dust surface PM concentration and total deposition, in the control and vegetation-impact simulations, using the 1-mode and 4-mode configurations, respectively. Statistics for DAOD, surface PM, and deposition are calculated only for grid cells where the relative difference between control and vegetation-impact simulations exceeds 3 % (or 5 %), consistent with the thresholds applied in Figs. 5, 6, 8, 10, S20, and S25.

Variables	Size-mode	Simulation	R^2	R	RMSE	NRMSE	MB
Emission against DustCOMM by Kok et al., 2021b	1-mode	Control	0.87	0.93	57.58	42 %	39.86
		Vegetation-impact	1.00	1.00	0.18	0 %	0.01
	4-mode	Control	0.86	0.93	551.15	43 %	387.09
		Vegetation-impact	0.99	1.00	94.57	7 %	0.13
DAOD against AERONET-SDA	1-mode	Control	0.68	0.82	0.09	66 %	0.06
		Vegetation-impact	0.63	0.80	0.07	53 %	0.03
	4-mode	Control	0.67	0.82	0.10	71 %	0.06
		Vegetation-impact	0.68	0.83	0.07	49 %	0.03
DAOD against model constraints by Kok et al., 2021a	1-mode	Control	0.52	0.72	0.10	73 %	0.02
		Vegetation-impact	0.57	0.76	0.09	60 %	0.00
	4-mode	Control	0.47	0.68	0.12	83 %	0.02
		Vegetation-impact	0.53	0.73	0.09	65 %	0.00
Surface PM concentration against Mahowald et al., 2009	1-mode	Control	0.81 (log space)	0.90 (log space)	0.44 (log space)	10 % (log space)	-0.02 (log space)
		Vegetation-impact	0.82 (log space)	0.91 (log space)	0.45 (log space)	10 % (log space)	-0.12 (log space)
	4-mode	Control	0.83 (log space)	0.91 (log space)	0.51 (log space)	12 % (log space)	0.15 (log space)
		Vegetation-impact	0.84 (log space)	0.92 (log space)	0.49 (log space)	12 % (log space)	0.03 (log space)
Surface deposition against Albani et al. (2014)	1-mode	Control	0.66 (log space)	0.81 (log space)	0.80 (log space)	14 % (log space)	0.15 (log space)
		Vegetation-impact	0.70 (log space)	0.84 (log space)	0.74 (log space)	13 % (log space)	-0.01 (log space)
	4-mode	Control	0.59 (log space)	0.77 (log space)	0.96 (log space)	19 % (log space)	0.36 (log space)
		Vegetation-impact	0.64 (log space)	0.80 (log space)	0.87 (log space)	17 % (log space)	0.19 (log space)

Table S4. Global and regional dust emissions and their relative contributions under control and vegetation-impact simulations in the 1-mode configuration.

Regions	Dust emission (Tg y^{-1})			Contribution of each region (%)	
	Control simulation	Vegetation-impact simulation	Relative reduction (%)	Control simulation	Vegetation-impact simulation
Global	1601	1243	22 %	——	
Western North Africa	234	233	1 %	15 %	19 %
Eastern North Africa	193	190	1 %	12 %	15 %
The Middle East*	166	155	7 %	10 %	13 %
East Asia	168	154	8 %	10 %	12 %
The Sahel	168	148	12 %	11 %	12 %
Central Asia*	298	210	30 %	19 %	17 %
Australia	64	42	34 %	4 %	3 %
South America	89	50	44 %	6 %	4 %
Southern Africa	59	27	55 %	4 %	2 %
North America	162	34	79 %	10 %	3 %

*Note: To provide a more detailed regional characterization, the Middle East and Central Asia (originally aggregated in Kok et al. (2021a), Fig. S4) were differentiated: The Middle East is defined as the region situated south of 40°N and west of 60°E, and the remaining areas of the original domain, are categorized as Central Asia.

Table S5. Global and regional dust emissions, DAOD, surface concentrations, and deposition rates for the control and vegetation-impact simulations in the 1-mode configuration. Relative reductions (%) between the two simulations are also provided.

	Dust emission (Tg y^{-1})			DAOD			Surface PM concentration (μg m^{-3})			Surface total deposition (g m^{-2} y^{-1})		
	Control	Vegetation-impact	Relative reduction (%)	Control	Vegetation-impact	Relative reduction (%)	Control	Vegetation-impact	Relative reduction (%)	Control	Vegetation-impact	Relative reduction (%)
Global	1601	1243	22 %	0.031	0.025	19 %	12	10	17 %	3	2	23 %
Western North Africa	234	233	1 %	0.249	0.238	4 %	117	115	2 %	20	19	4 %
Eastern North Africa	193	190	1 %	0.204	0.194	5 %	94	92	2 %	17	16	2 %
The Middle East*	166	155	7 %	0.155	0.139	11 %	63	59	7 %	14	13	8 %
East Asia	168	154	8 %	0.137	0.121	12 %	68	62	10 %	15	13	9 %
The Sahel	168	148	12 %	0.222	0.209	6 %	90	85	6 %	19	18	7 %
Central Asia*	298	210	30 %	0.201	0.160	20 %	90	69	23 %	26	23	13 %
Australia	64	42	34 %	0.026	0.017	34 %	15	10	33 %	3	2	34 %
South America	89	50	44 %	0.012	0.008	37 %	3	2	46 %	3	2	43 %
Southern Africa	59	27	55 %	0.026	0.015	43 %	10	5	47 %	4	2	52 %
North America	162	34	79 %	0.042	0.013	69 %	21	5	75 %	7	2	75 %

*Note: To provide a more detailed regional characterization, the Middle East and central Asia (originally aggregated in Kok et al. (2021a), Fig. S4) were differentiated: The Middle East is defined as the region situated south of 40°N and west of 60°E, and the remaining areas of the original domain, are categorized as central Asia.

Table S6. Global and regional dust emissions, DAOD, surface concentrations, and deposition rates for the control and vegetation-impact simulations in the 4-mode configuration. Relative reductions (%) between the two simulations are also provided.

	Dust emission (Tg y^{-1})			DAOD			Surface PM_{10} concentration (μg m^{-3})			Surface PM_{10} total deposition (g m^{-2} y^{-1})		
	Control	Vegetation-impact	Relative reduction (%)	Control	Vegetation-impact	Relative reduction (%)	Control	Vegetation-impact	Relative reduction (%)	Control	Vegetation-impact	Relative reduction (%)
Global	15058	11575	23 %	0.031	0.025	19 %	25	20	19 %	12	9	23 %
Western North Africa	2098	2082	1 %	0.255	0.247	3 %	254	251	1 %	112	108	4 %
Eastern North Africa	1768	1744	1 %	0.208	0.201	4 %	207	203	2 %	118	116	1 %
The Middle East*	1467	1363	7 %	0.148	0.133	10 %	121	112	7 %	70	65	7 %
East Asia	1556	1433	8 %	0.147	0.132	11 %	156	142	9 %	61	57	6 %
The Sahel	1518	1344	11 %	0.222	0.209	6 %	179	167	6 %	64	58	9 %
Central Asia*	2887	1957	32 %	0.214	0.168	21 %	207	153	26 %	115	96	16 %
Australia	626	409	35 %	0.028	0.019	35 %	32	21	34 %	14	9	35 %
South America	1263	720	43 %	0.020	0.012	40 %	13	7	47 %	26	15	42 %
Southern Africa	556	229	59 %	0.026	0.013	49 %	22	11	51 %	24	10	57 %
North America	1320	295	78 %	0.042	0.012	71 %	46	12	75 %	35	8	76 %

*Note: To provide a more detailed regional characterization, the Middle East and central Asia (originally aggregated in Kok et al. (2021a), Fig. S4) were differentiated: The Middle East is defined as the region situated south of 40°N and west of 60°E, and the remaining areas of the original domain, are categorized as central Asia.

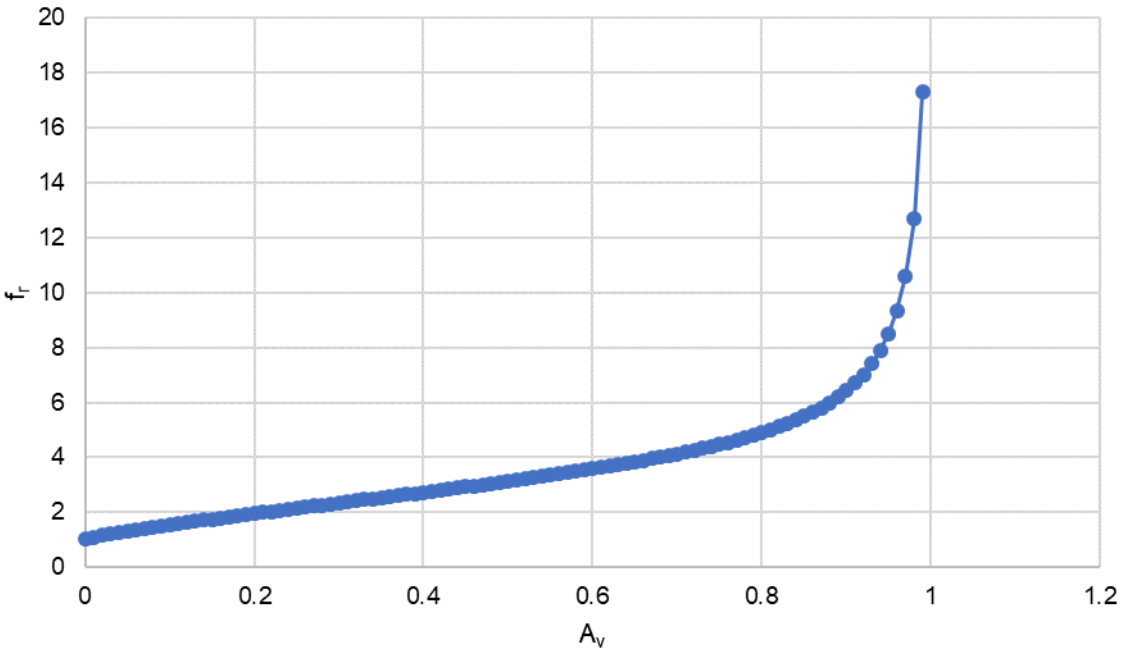

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Figure S1. Relationship between the fraction of vegetation cover (A_v) and the correction factor related to surface roughness (f_r).

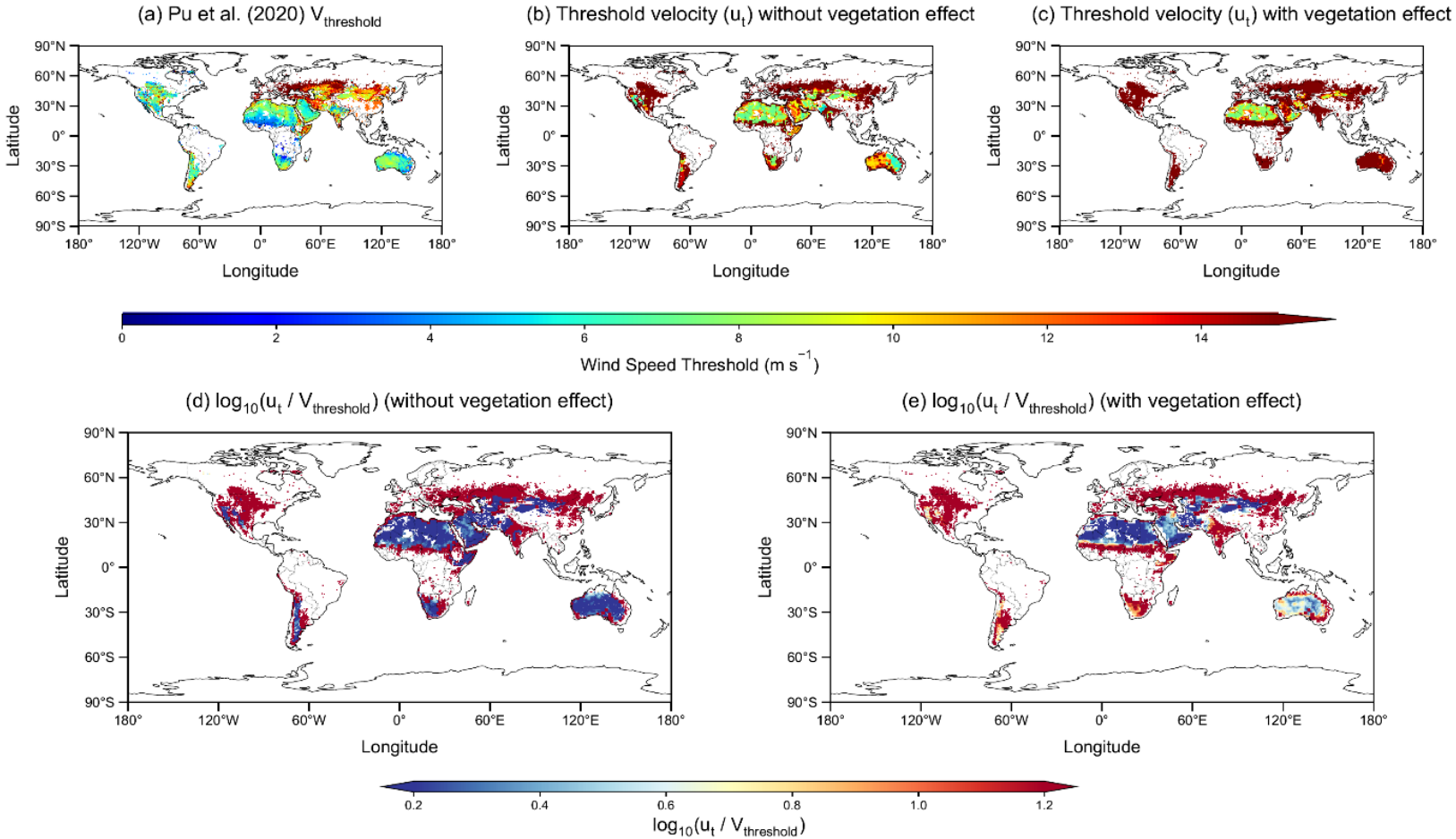


Figure S2. Comparison between observationally-retrieved and modelled dust emission threshold velocities. (**a**) Satellite-retrieved 10-m wind speed threshold ($V_{threshold}$) from Pu et al. (2020), using a high dust intensity threshold (DOD > 0.5) to identify active emission events. (**b**, **c**) Model-simulated threshold velocity (u_t) without and with the effect of vegetation, respectively. (**d**, **e**) Spatial distribution of the logarithmic ratio, $\log_{10}(u_t/V_{threshold})$, excluding and including vegetation effects. All model data were interpolated to a 0.5° × 0.5° spatial resolution to match the observational grid.

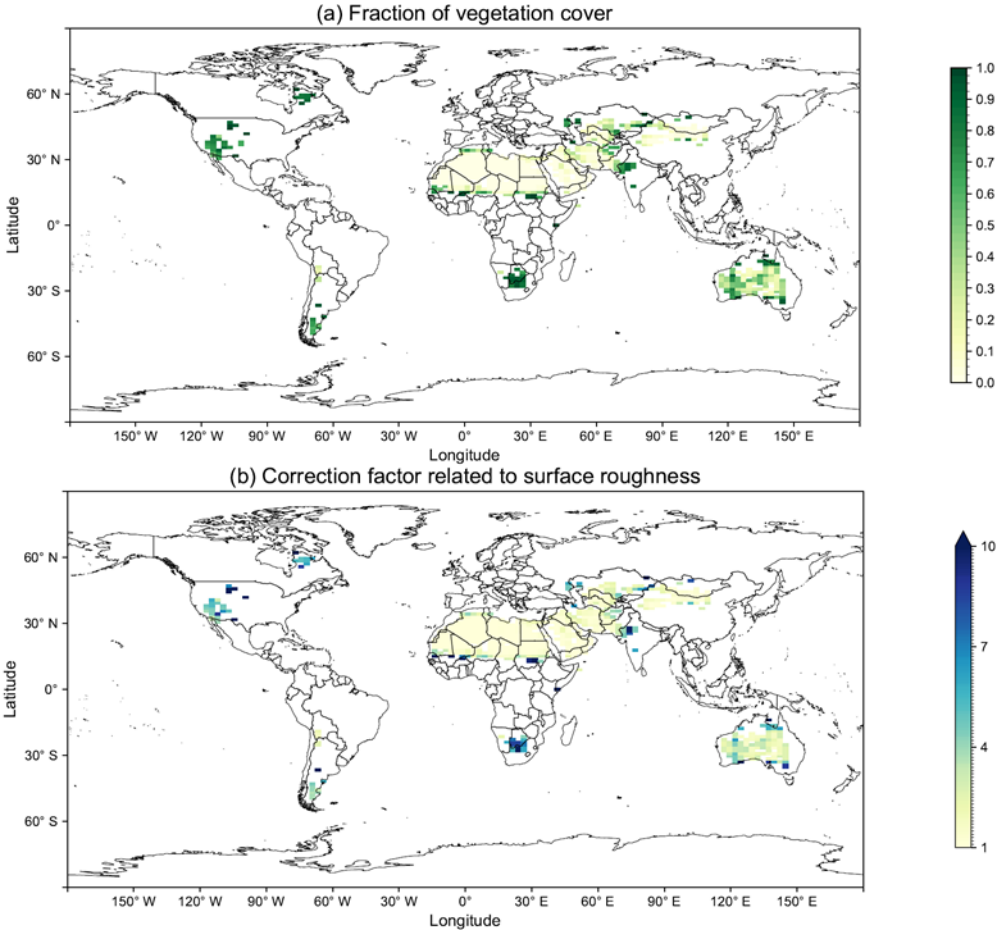


Figure S3. Global distribution of vegetation cover (A_v) and surface roughness correction factor (f_r), averaged over 2004–2020. (**a**) Fraction of vegetation cover (A_v). The A_v values, including the grasslands, forests, and croplands, are derived from the ORCHIDEE land surface model. (**b**) Correction factor related to surface roughness (f_r). The f_r values are parameterized based on the fraction of vegetation cover (A_v). In both panels, white color represents regions without valid data (e.g., oceans) or where the values fall outside the defined display range (e.g., f_r greater than 100). Both fields represent 17-year averages (2004–2020) and are shown only for grid cells where dust emission is permitted.

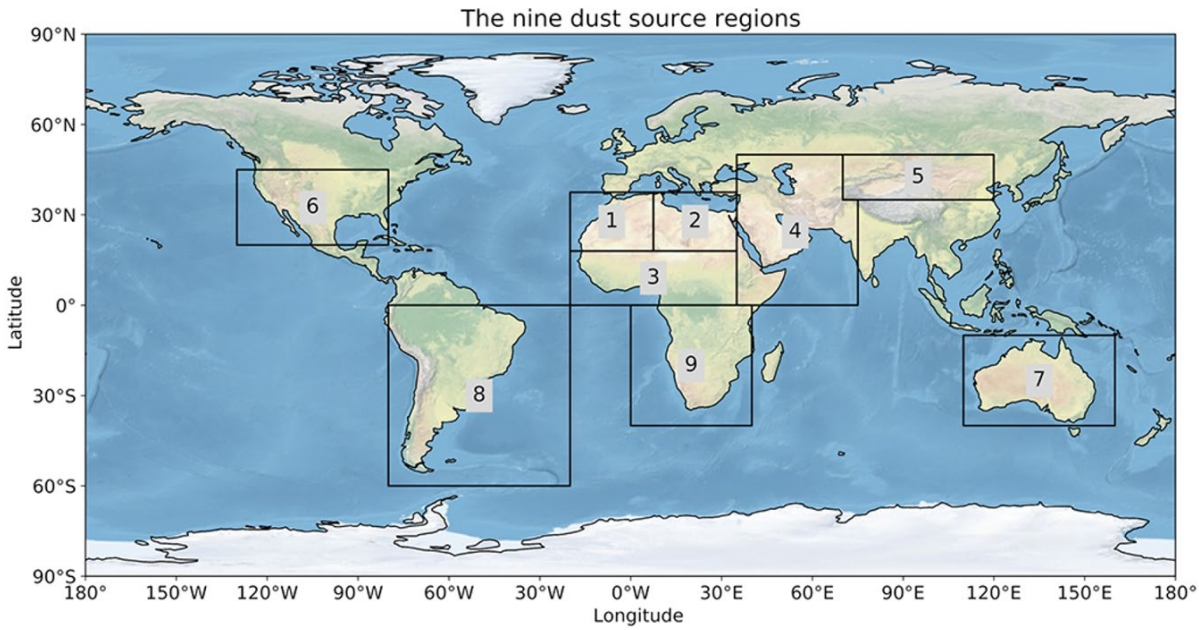


Figure S4. Geographical location of nine dust source regions. The nine regions include (1) western North Africa, (2) eastern North Africa, (3) the southern Sahara and Sahel, (4) the Middle East and central Asia (which includes the Horn of Africa), (5) East Asia, (6) North America, (7) Australia, (8) South America, and (9) southern Africa. The graph is taken from Kok et al. (2021a).

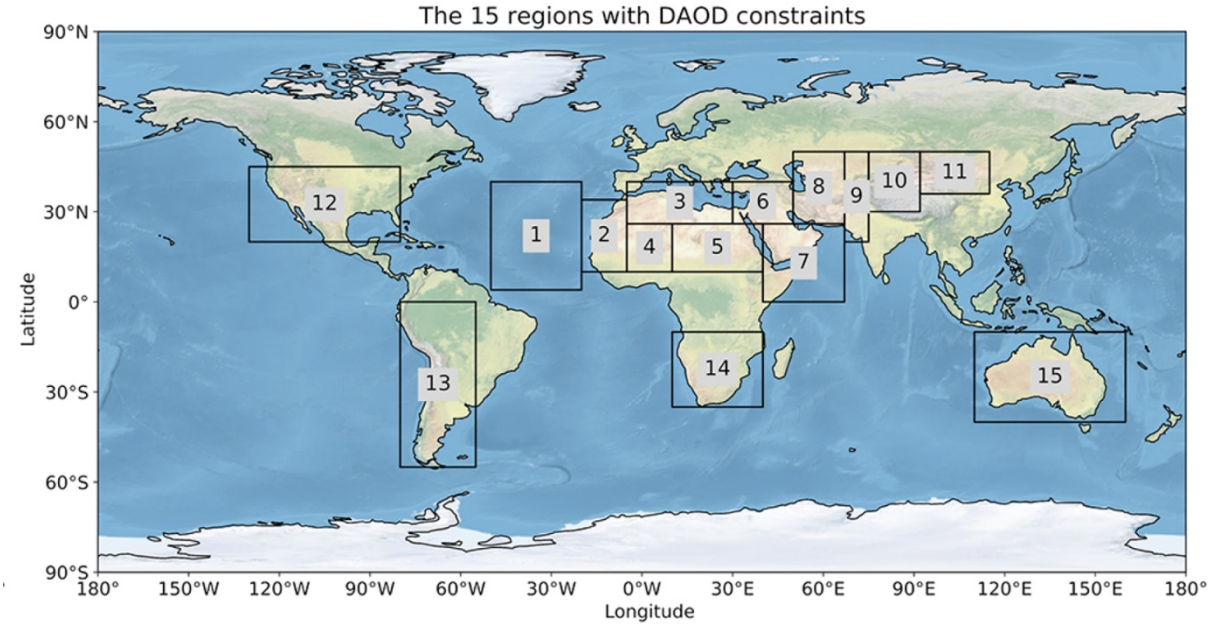


Figure S5. Geographical location of the 15 observed dusty regions. The regions include: (1) Mid-Atlantic, (2) African west coast, (3) Northern Africa, (4) Mali /Niger, (5) Bodélé / Sudan, (6) Northern Middle East, (7) Southern Middle East, (8) Kyzylkum, (9) Thar, (10) Taklamakan, (11) Gobi, (12) North America, (13) South America, (14) southern Africa, and (15) Australia. The graph is taken from Kok et al. (2021a).

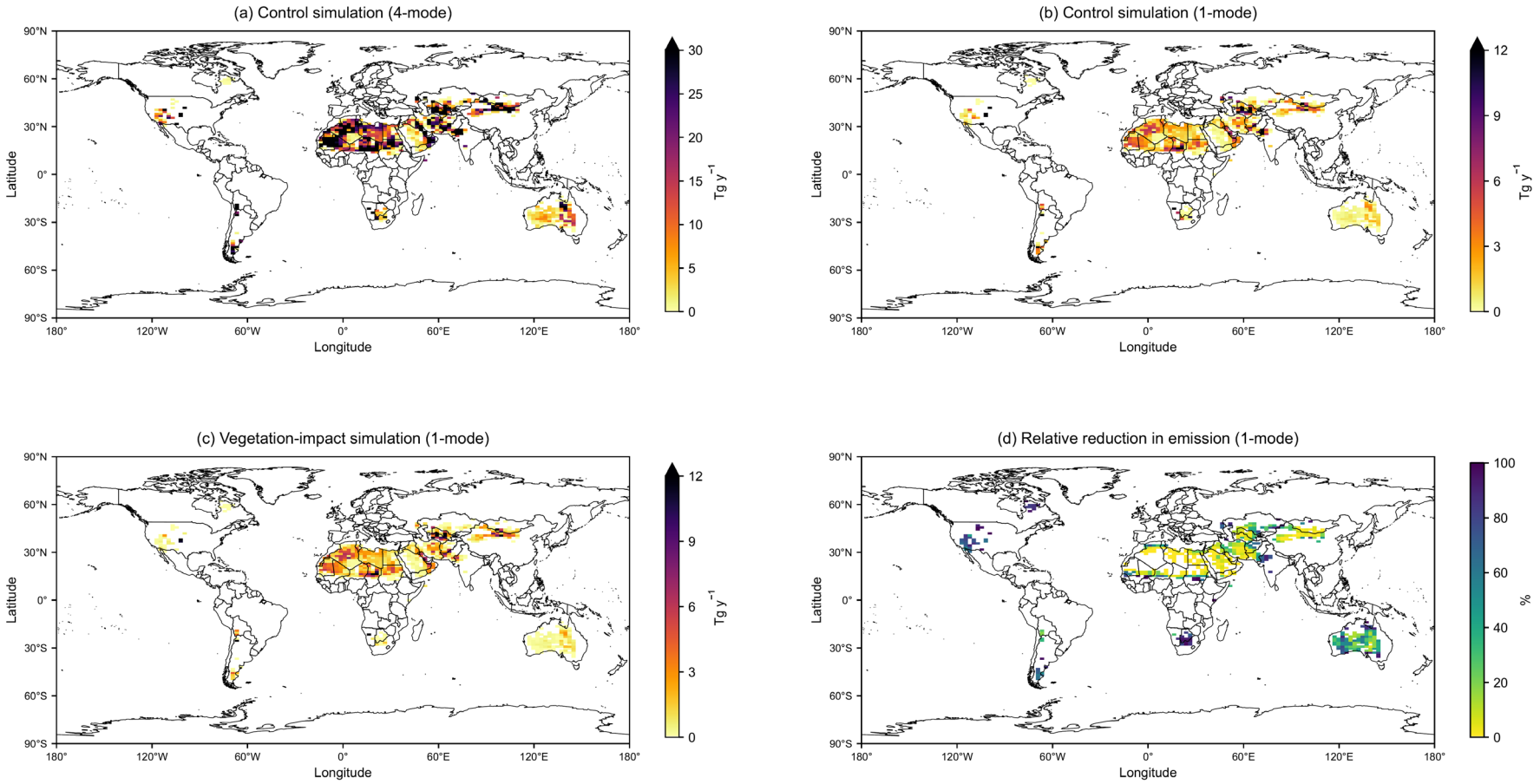


Figure S6. Simulated global dust emission fluxes and the quantified impact of vegetation. Mean annual dust emission flux (Tg y^{-1}) in (**a**) the control simulation of the 4-mode configuration, and in (**b**, **c**) the control and vegetation-impact simulations of the 1-mode configuration, respectively. (**d**) Relative reduction in dust emission (%) due to vegetation in the 1-mode configuration, calculated as (control − vegetation-impact) / control.

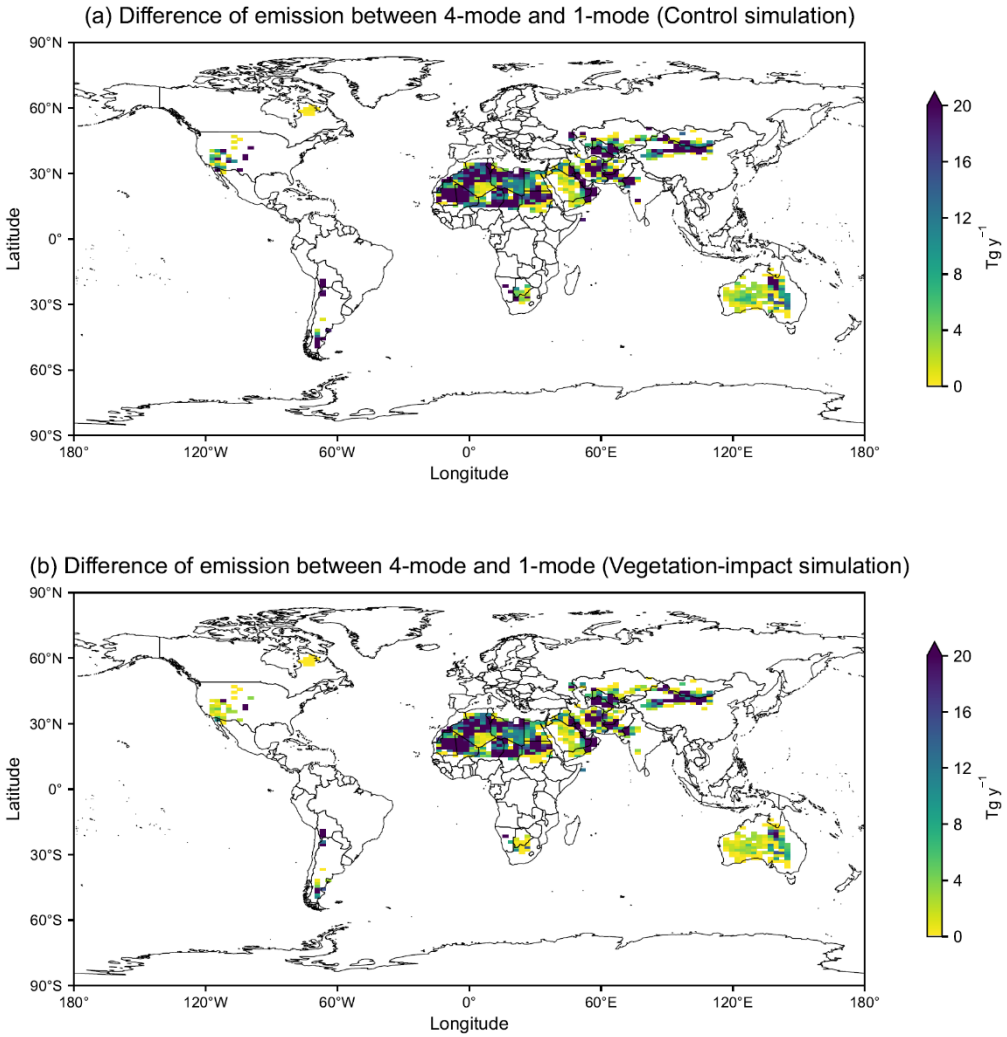


Figure S7. Difference in mean annual dust emission between the 4-mode and 1-mode configurations in the (**a**) control simulation and the (**b**) vegetation-impact simulation, calculated as 4-mode minus 1-mode.

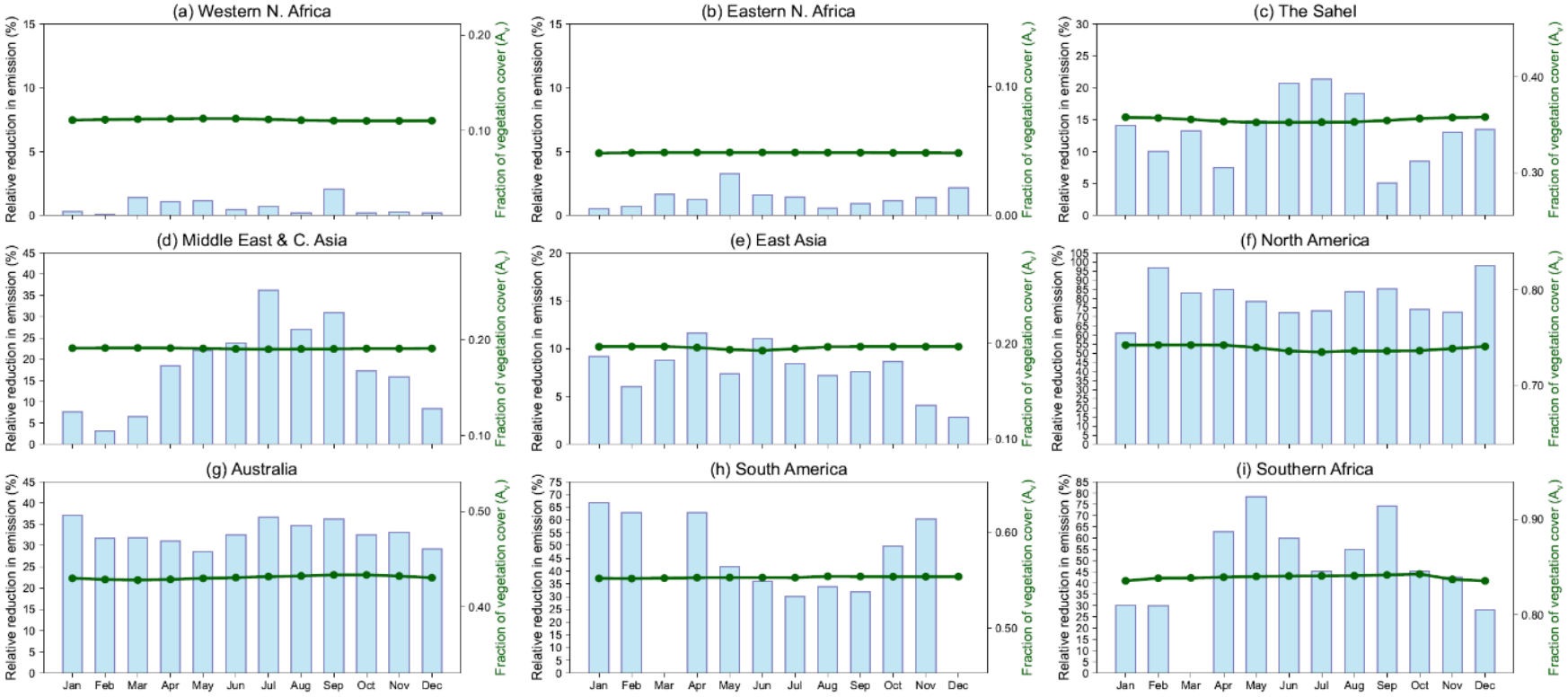
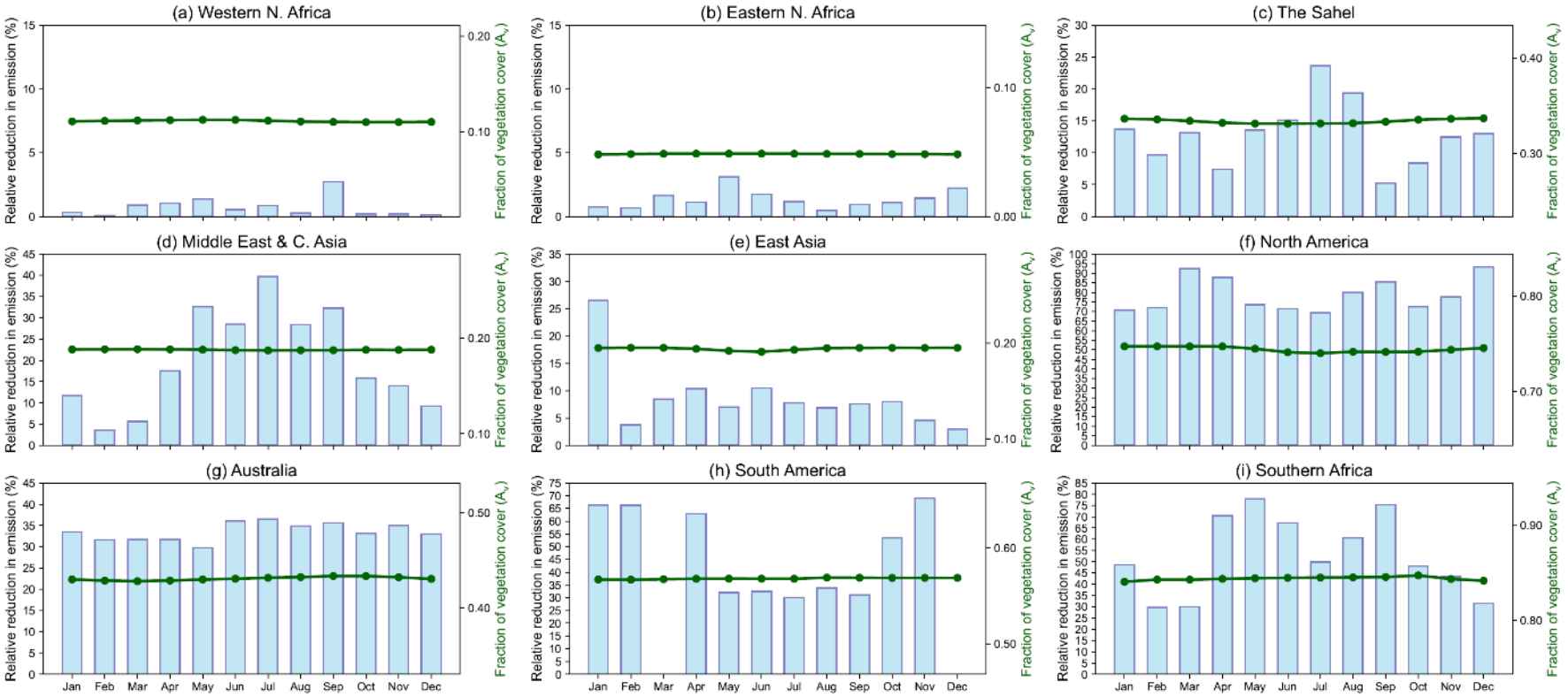


Figure S8. Seasonality of relative reduction in dust emission and fraction of vegetation cover (A_v) in the 1-mode configuration. Monthly variations are shown for nine dust-source regions: (**a**) Western North Africa, (**b**) Eastern North Africa, (**c**) the Sahel, (**d**) the Middle East and central Asia, (**e**) East Asia, (**f**) North America, (**g**) Australia, (**h**) South America, and (**i**) southern Africa. The left y-axis represents the relative reduction in emission (%), calculated as (Control−Vegetation-impact)/Control. The right y-axis represents the fraction of vegetation cover (A_v). Bars denote emission reduction, and the line with markers denotes the vegetation fraction.



185 **Figure S9.** Seasonality of relative reduction in dust emission and fraction of vegetation cover (A_v) in the 4-mode configuration. Monthly variations are shown for nine dust-source regions: (**a**) Western North Africa, (**b**) Eastern North Africa, (**c**) the Sahel, (**d**) the Middle East and central Asia, (**e**) East Asia, (**f**) North America, (**g**) Australia, (**h**) South America, and (**i**) southern Africa. The left y-axis represents the relative reduction in emission (%), calculated as (Control−Vegetation-impact)/Control. The right y-axis represents the fraction of vegetation cover (A_v). Bars denote emission reduction, and the line with markers denotes the vegetation fraction.

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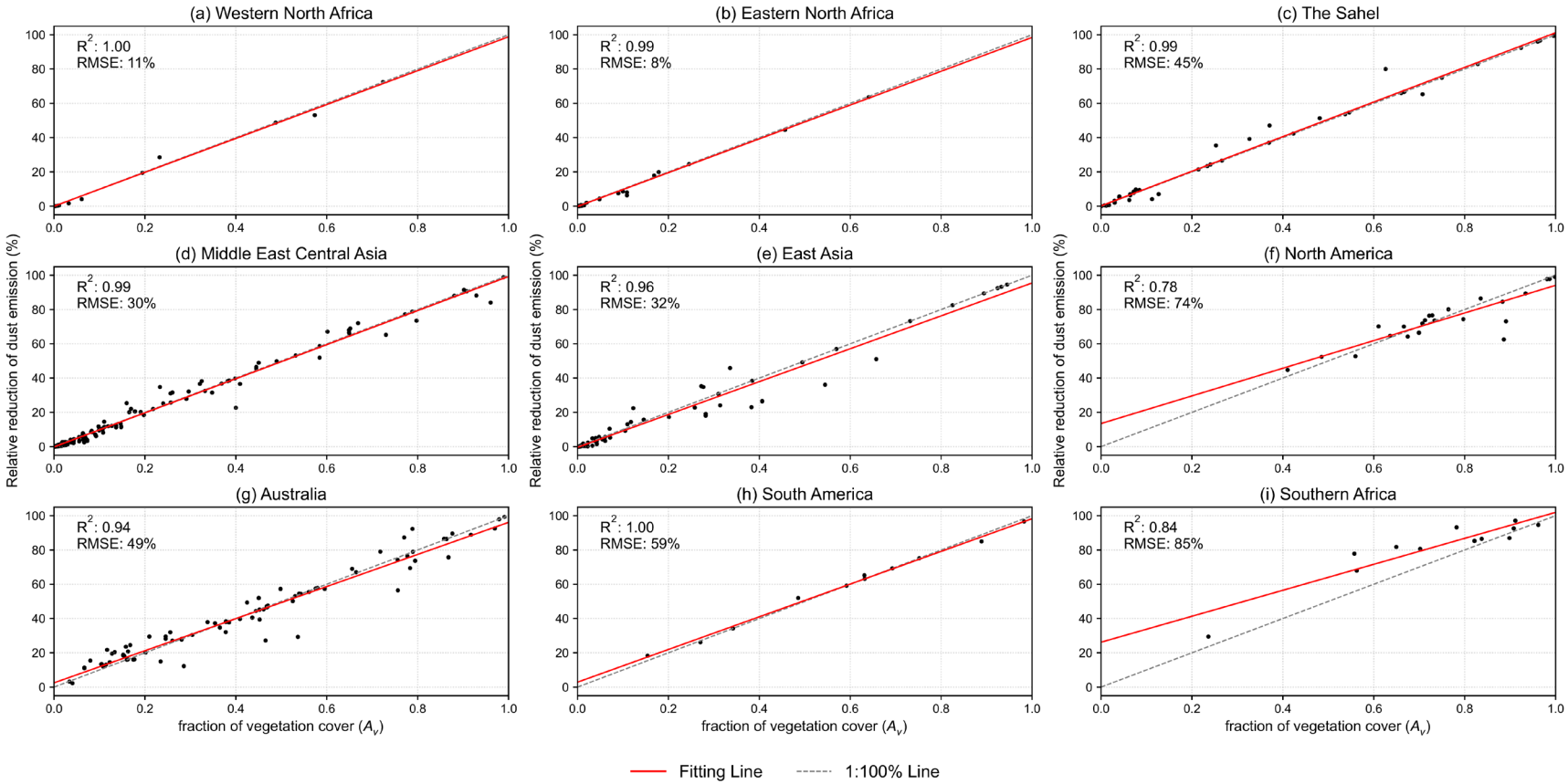

Figure S10. Relationship between the mean annual fraction of vegetation cover (A_v) and the relative reduction of dust emission in 1-mode configuration for nine dust source regions, including (**a**) Western North Africa, (**b**) Eastern North Africa, (**c**) the Sahel, (**d**) the Middle East and central Asia, (**e**) East Asia, (**f**) North America, (**g**) Australia, (**h**) South America, and (**i**) southern Africa. Each point represents a grid cell. The relative reduction is calculated as (control − vegetation-impact) / control.

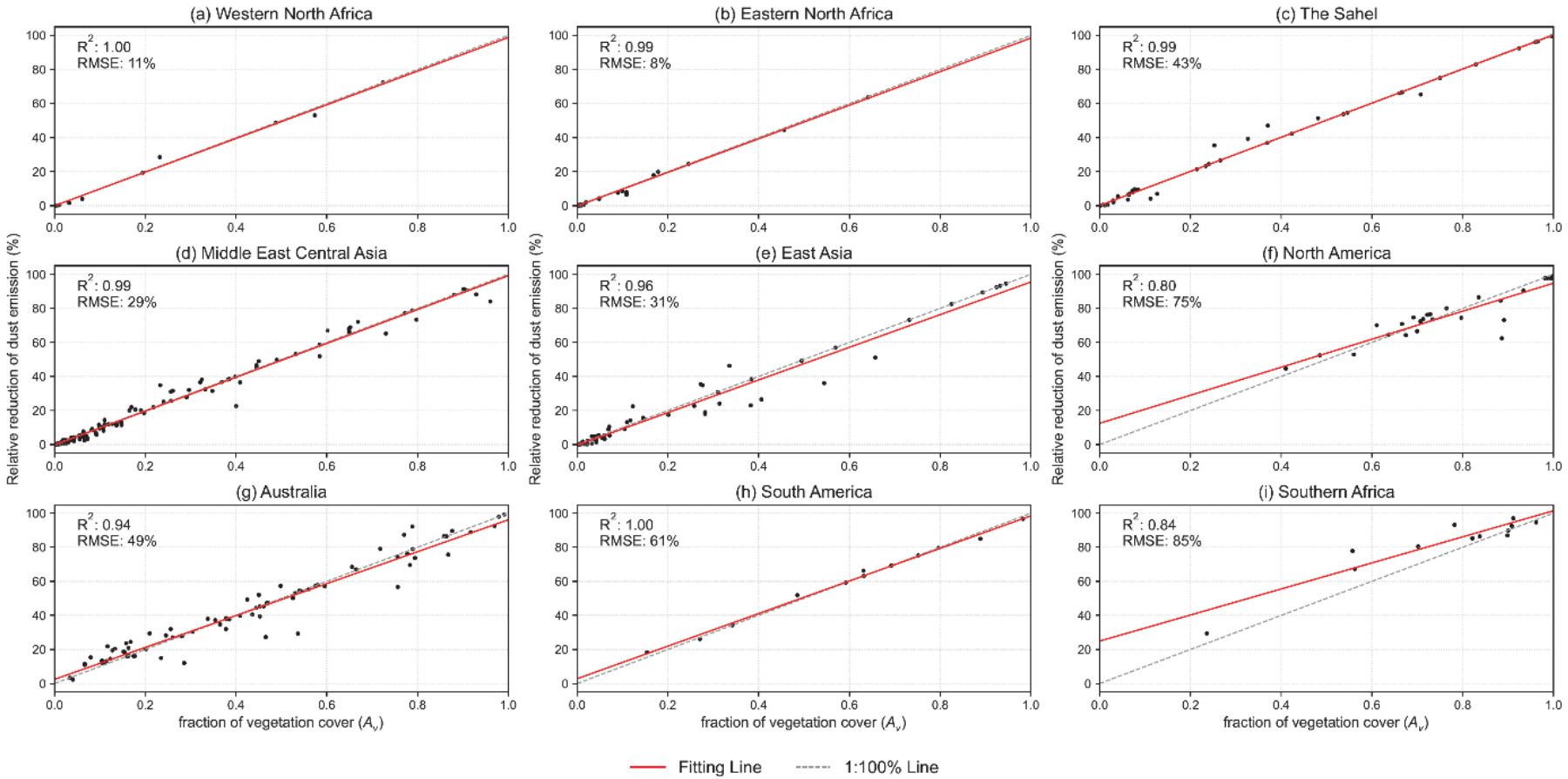

Figure S11. Relationship between the mean annual fraction of vegetation cover (A_V) and the relative reduction of dust emission in 4-mode configuration for nine dust source regions, including (**a**) Western North Africa, (**b**) Eastern North Africa, (**c**) the Sahel, (**d**) the Middle East and central Asia, (**e**) East Asia, (**f**) North America, (**g**) Australia, (**h**) South America, and (**i**) southern Africa. Each point represents a grid cell. The relative reduction is calculated as (control − vegetation-impact) / control.

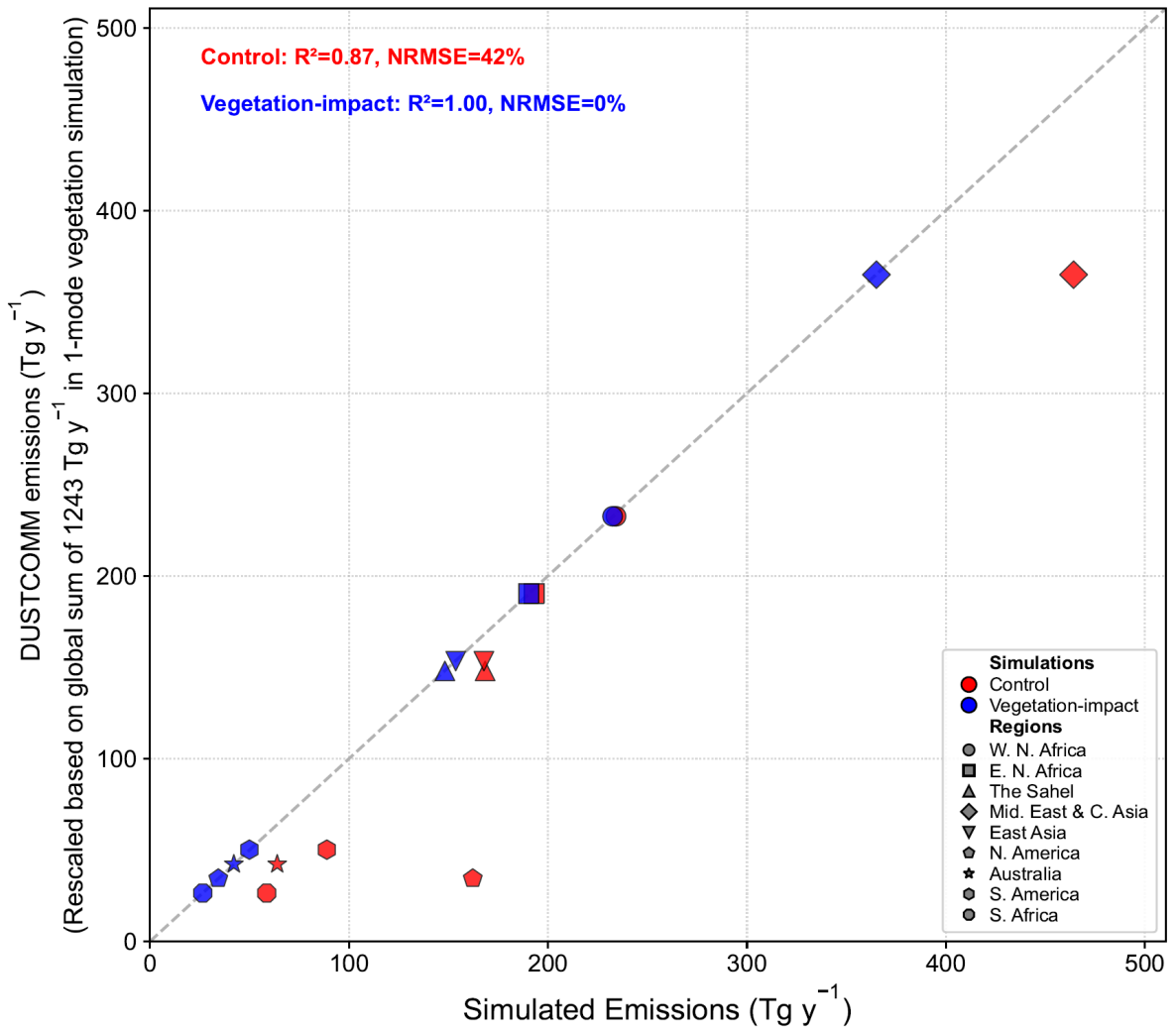

Figure S12. Comparison of simulated regional dust emissions (X-axis) against the DustCOMM emission by Kok et al. (2021b) (Y-axis) in 1-mode configuration. The rescaling factors were applied to both the control (red) and vegetation-impact (blue) simulations. The DustCOMM emissions were rescaled to the global total emissions in the vegetation-impact simulation. The 1:1 line, R^2, and NRMSE are provided for reference.

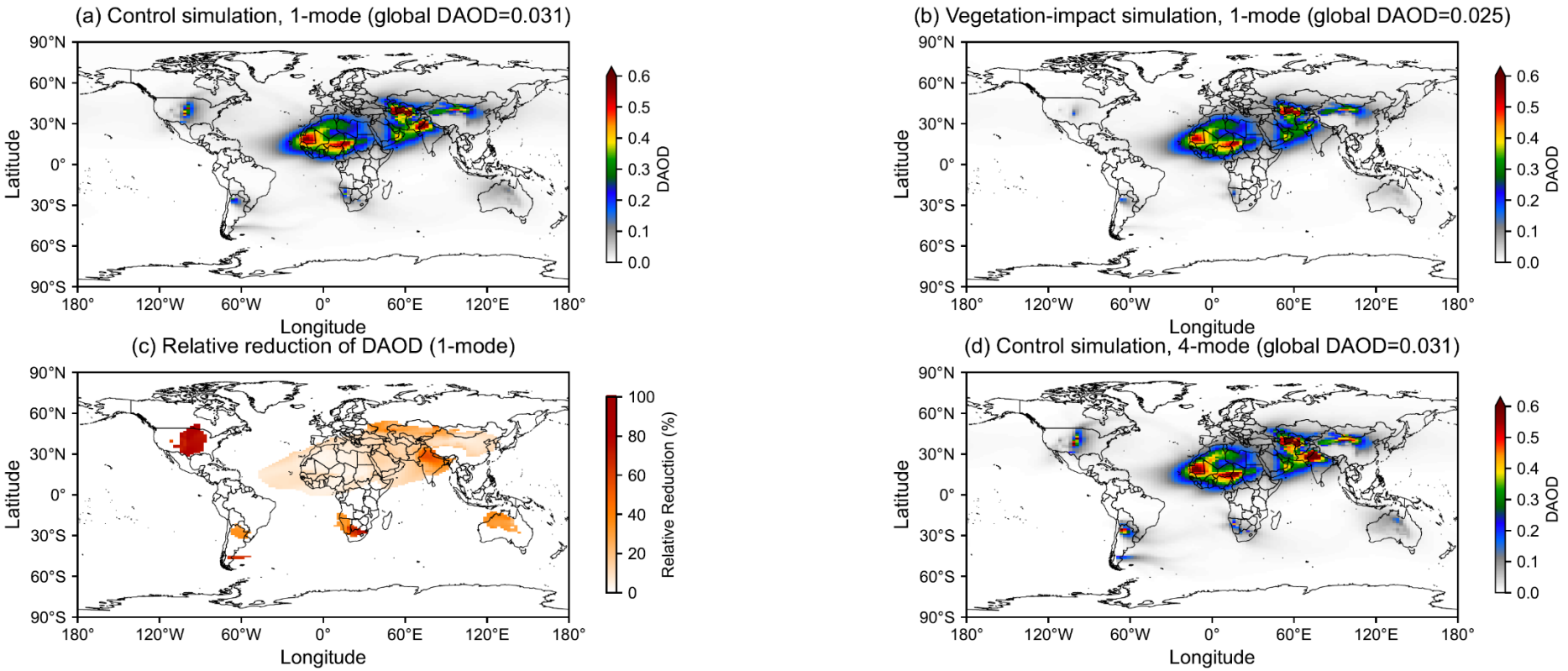

Figure S13. Spatial distribution of mean annual dust aerosol optical depth (DAOD) at 550 nm. Panels (**a**) and (**b**) show the control and vegetation-impact simulations using the 1-mode configuration, respectively. Panel (**c**) shows the relative difference between these two 1-mode simulations, calculated as (control − vegetation-impact) / control, with only grid cells with DAOD ≥ 0.05 displayed. Panel (**d**) presents the control simulation using the 4-mode configuration. The specific area-weighted global mean annual DAOD is indicated in panels **a**, **b**, **d**. All simulations incorporate scaling coefficients to match the observation-based target range of 0.030 ± 0.005.

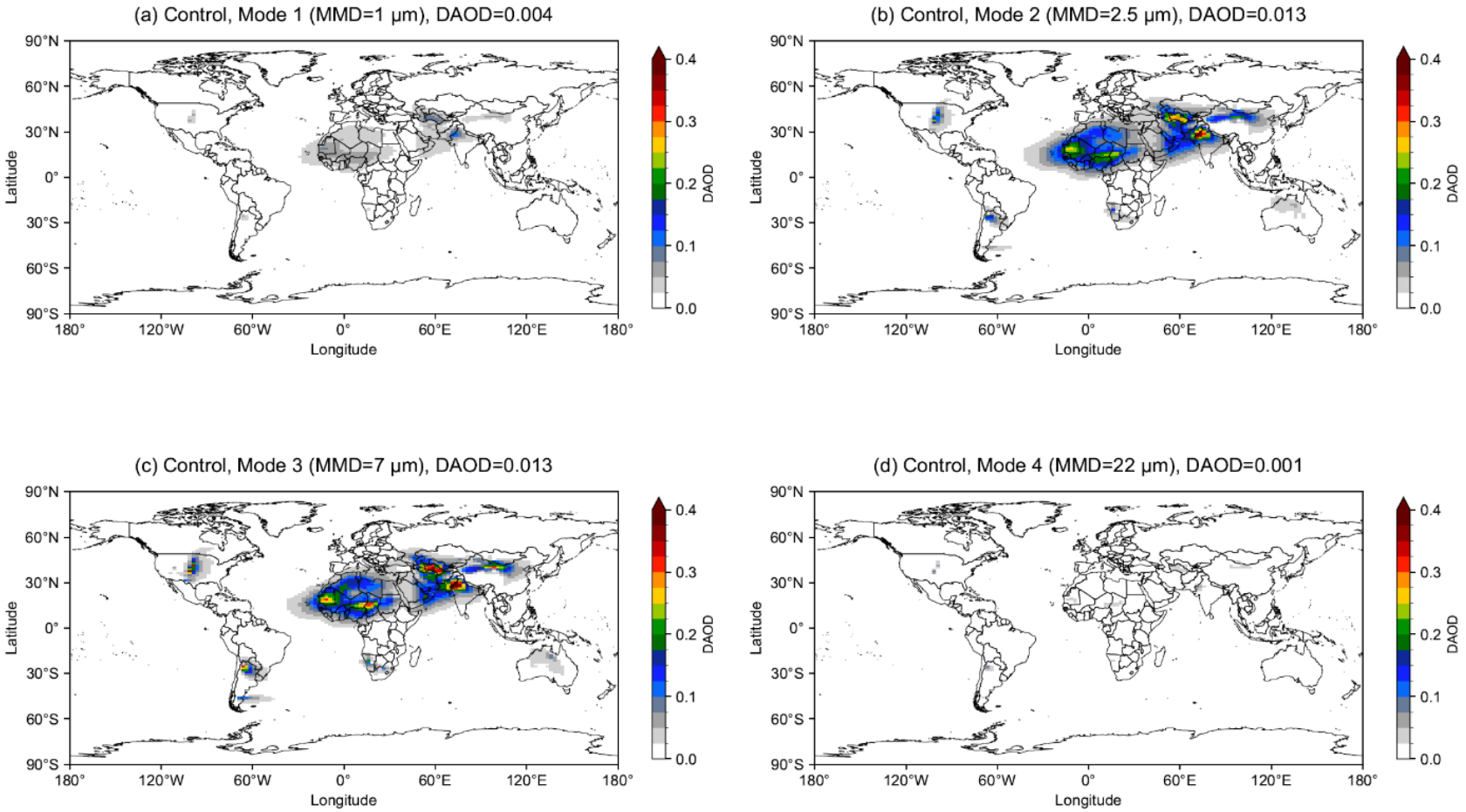

Figure S14. Mean annual Dust Aerosol Optical Depth (DAOD) at 550 nm for individual dust modes in the control simulation in four-mode configuration. Panels display: (**a**) Mode 1 (1 μm), (**b**) Mode 2 (2.5 μm), (**c**) Mode 3 (7 μm), and (**d**) Mode 4 (22 μm). The scaling coefficients were applied to align with the observation-based target range of 0.030 ± 0.005. The specific area-weighted global mean annual DAOD for each mode is indicated in the respective panel titles. All simulations incorporate scaling coefficients to match the observation-based target range of 0.030 ± 0.005.

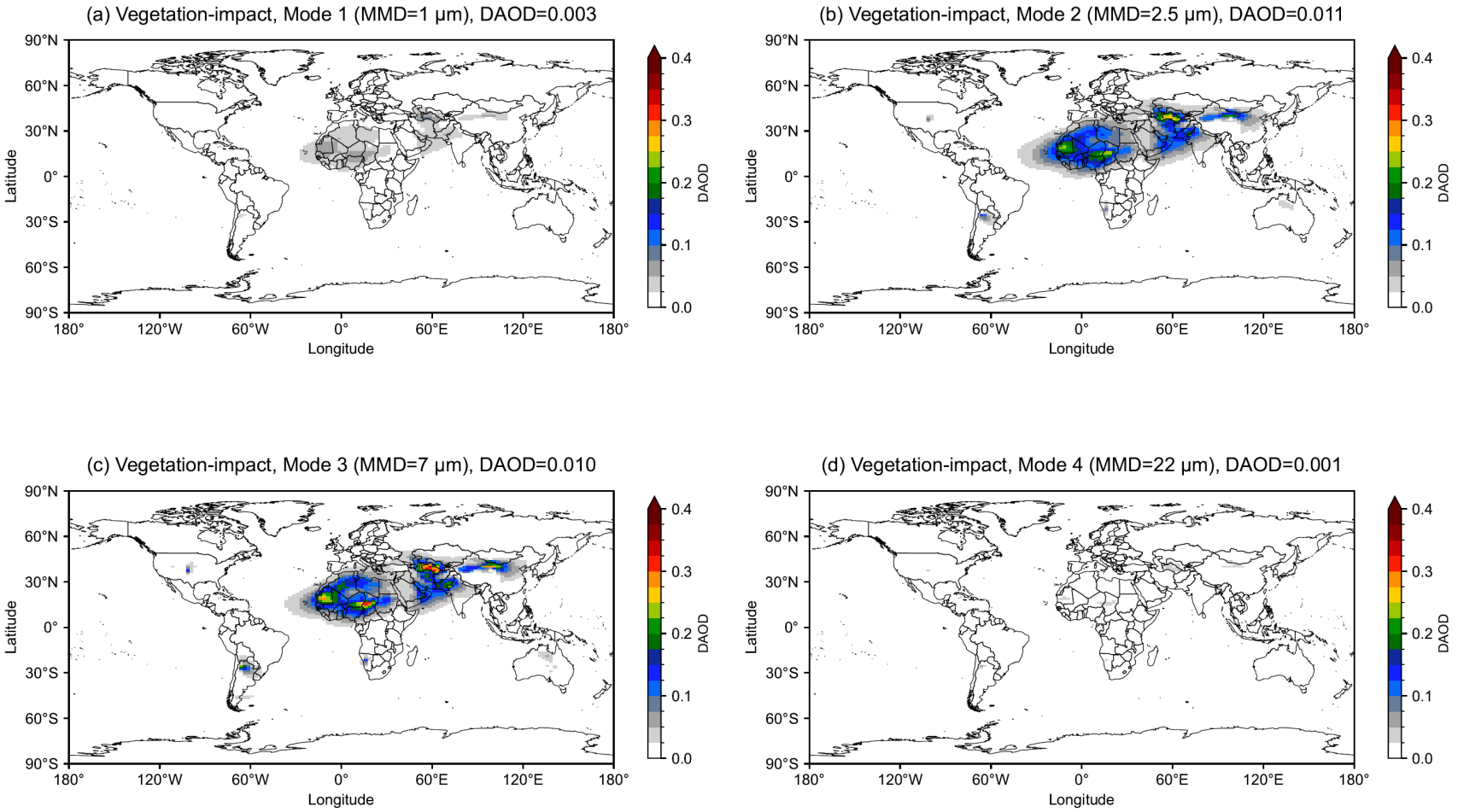

Figure S15. Mean annual Dust Aerosol Optical Depth (DAOD) at 550 nm for individual dust modes in the vegetation-impact simulation in four-mode configuration. Panels display: (**a**) Mode 1 (1 μm), (**b**) Mode 2 (2.5 μm), (**c**) Mode 3 (7 μm), and (**d**) Mode 4 (22 μm). The scaling coefficients were applied to align with the observation-based target range of 0.030 ± 0.005. The specific area-weighted global mean annual DAOD for each mode is indicated in the respective panel titles. All simulations incorporate scaling coefficients to match the observation-based target range of 0.030 ± 0.005.

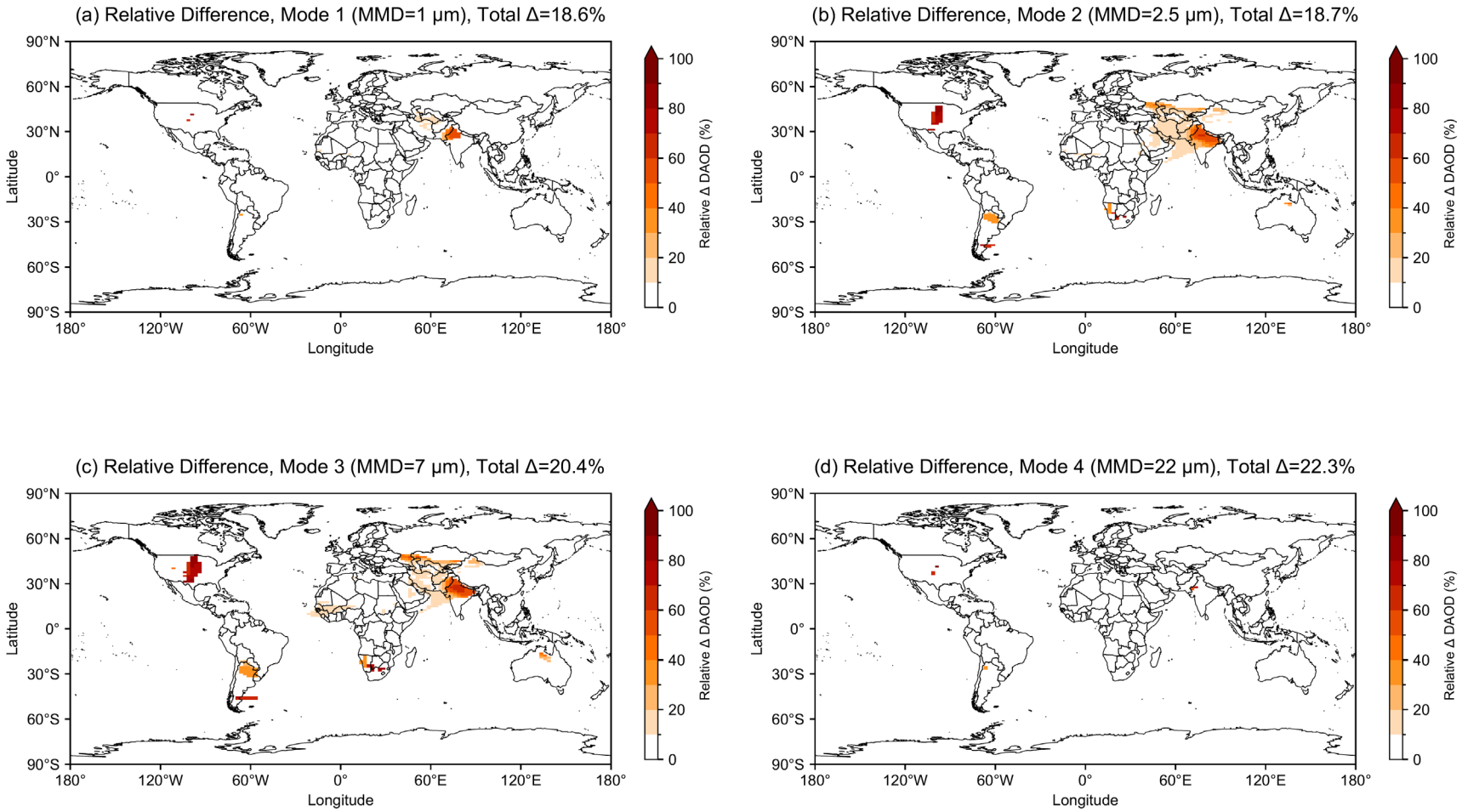

Figure S16. Relative difference of mean annual DAOD at 550 nm for individual dust modes between the control and the vegetation-impact simulation in four-mode configuration. Panels display: (**a**) Mode 1 (1 μm), (**b**) Mode 2 (2.5 μm), (**c**) Mode 3 (7 μm), and (**d**) Mode 4 (22 μm). The scaling coefficients were applied to align with the observation-based target range of 0.030 ± 0.005. Relative difference of DAOD is calculated as (control minus vegetation-impact) / control, with global total relative difference (Δ) for each mode provided in the respective sub-panel titles. Only grid cells with DAOD in control simulation ≥0.05 are shown. All simulations incorporate scaling coefficients to match the observation-based target range of 0.030 ± 0.005.

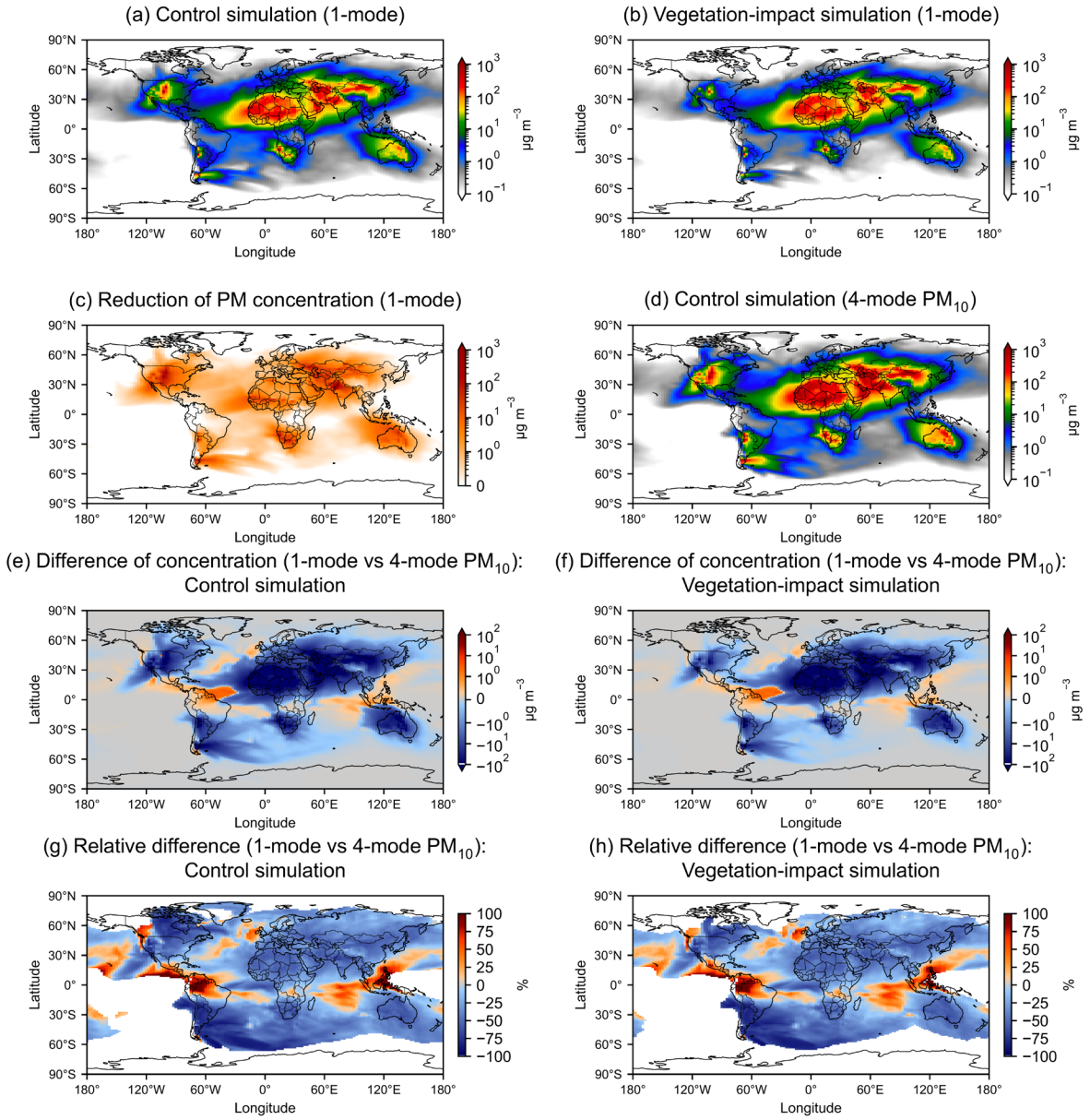

Figure S17. Spatial distribution of mean annual dust surface concentration (μg m^{-3}). Panels (**a**) and (**b**) show the control and vegetation-impact simulations using the 1-mode configuration, respectively. Panel (**c**) shows the reduction in surface concentration between these two 1-mode simulations, calculated as control − vegetation-impact. Panel (**d**) presents the dust surface PM_{10} concentration in the control simulation using the 4-mode configuration. Panels (**e**) and (**f**) show the absolute differences in surface concentration between the 1-mode and 4-mode (PM_{10} component) configurations for the control and vegetation-impact simulations, respectively, calculated as 1-mode − 4-mode PM_{10}. Panels (**g**) and (**h**) show the corresponding relative differences, calculated as (1-mode − 4-mode PM_{10}) / 1-mode.

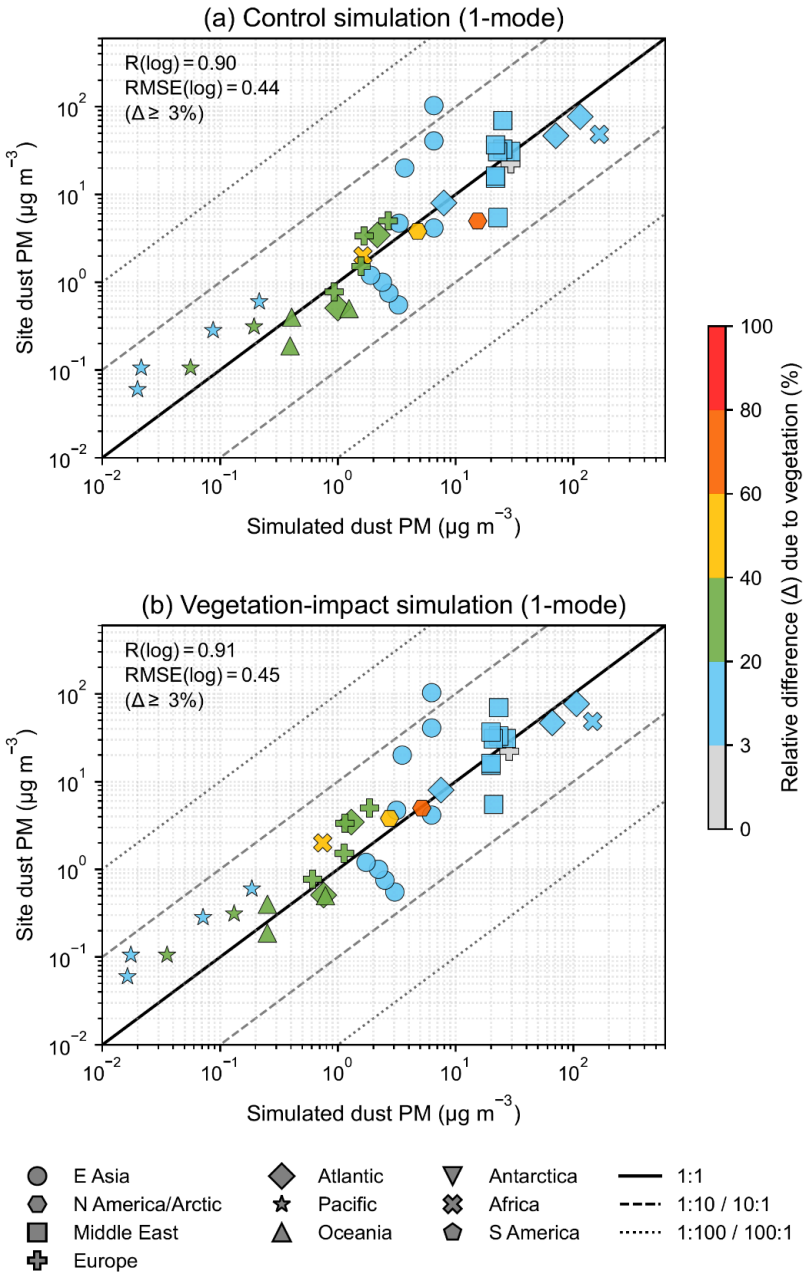

Figure S18. Evaluation of simulated and observed dust surface PM concentration (μg m^{-3}) by Mahowald et al. (2009) in the 1-mode configuration. Panels (**a**) and (**b**) show the control and vegetation-impact simulations, respectively. Both axes are plotted on logarithmic scales. Different regions are indicated by different marker shapes, with colours in each plot representing the relative difference between the control and vegetation-impact simulations. Statistical metrics (R and RMSE, computed in log space) are shown in each panel and are calculated only for data points with relative differences greater than 3 %, highlighting regions sensitive to vegetation impacts. The solid black line denotes the 1:1 relationship. The inner dashed lines indicate the 10:1 and 1:10 ratios, representing simulated values within one order of magnitude of the observations, while the outer dashed lines represent two orders of magnitude (100:1 and 1:100 ratios). Some observational sites are not visible because their values fall below the lower axis limits.

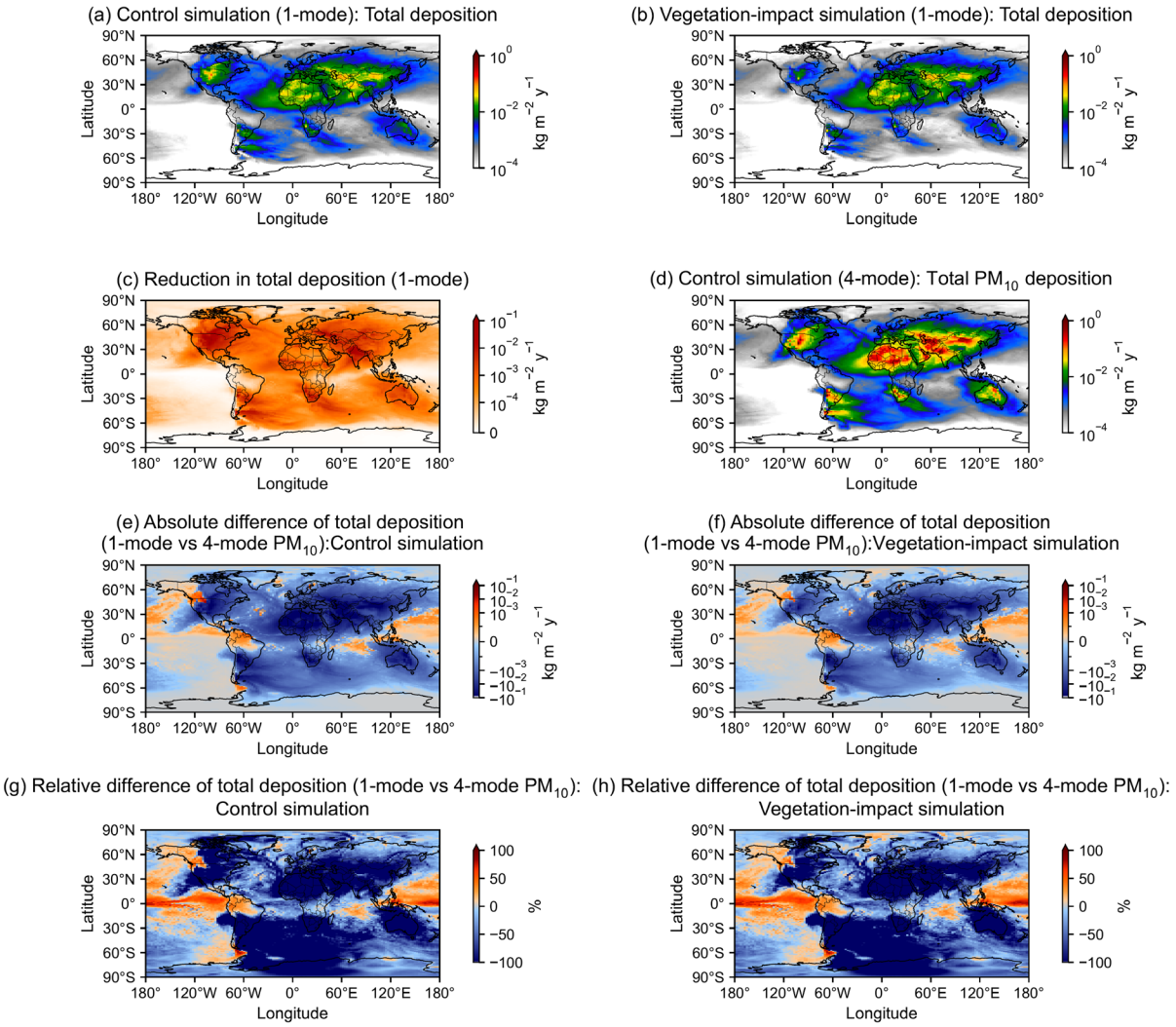

Figure S19. Spatial distribution of mean annual dust surface total deposition (kg m^{-2} y^{-1}). Panels (**a**) and (**b**) show the control and vegetation-impact simulations using the 1-mode configuration, respectively. Panel (**c**) shows the reduction in surface total deposition between these two 1-mode simulations, calculated as control − vegetation-impact. Panel (**d**) presents the dust surface PM_{10} total deposition in the control simulation using the 4-mode configuration. Panels (**e**) and (**f**) show the absolute differences in surface total deposition between the 1-mode and 4-mode (PM_{10} component) configurations for the control and vegetation-impact simulations, respectively, calculated as 1-mode − 4-mode PM_{10}. Panels (**g**) and (**h**) show the corresponding relative differences, calculated as (1-mode − 4-mode PM_{10}) / 1-mode.

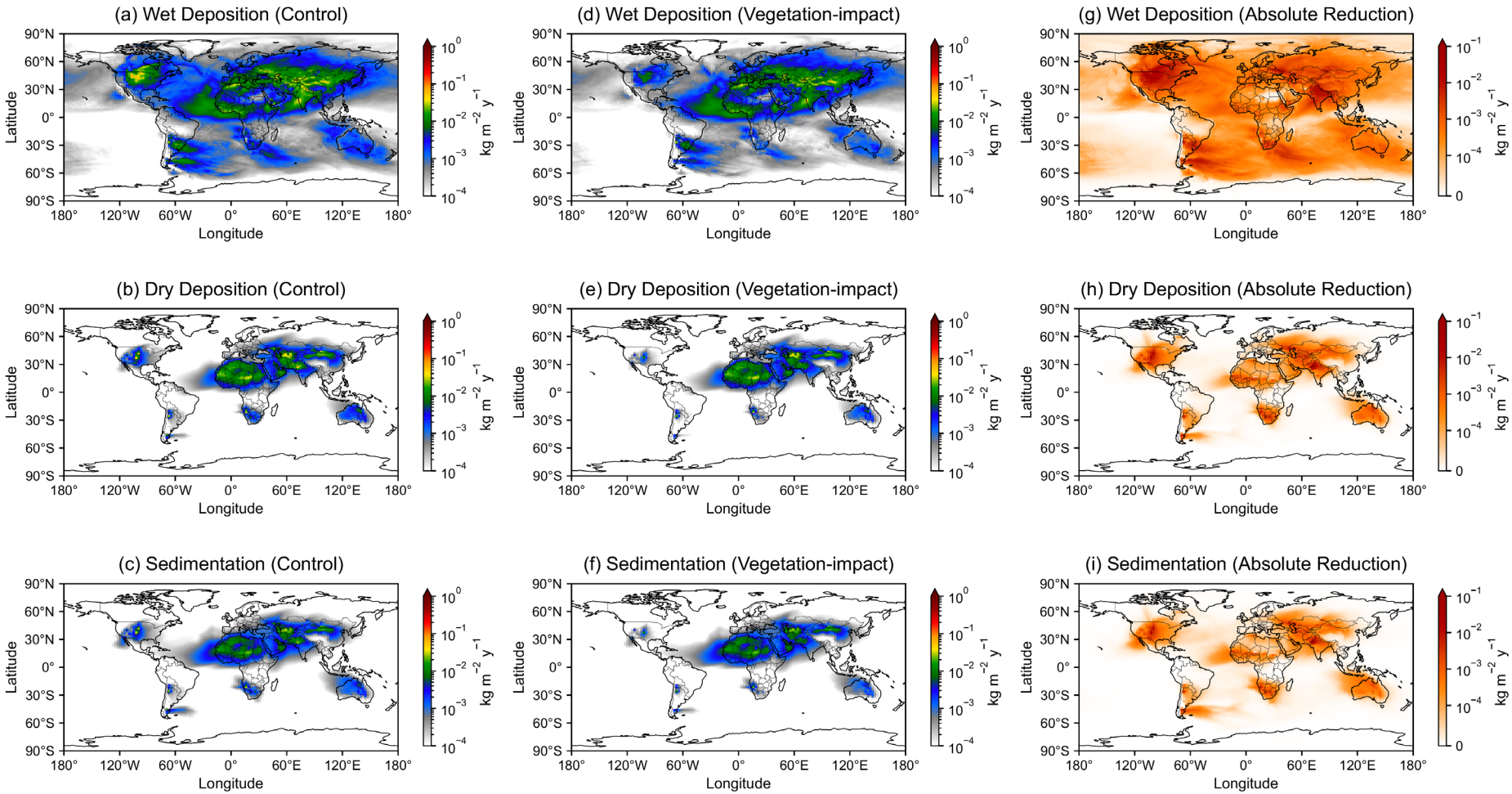

Figure S20. Global dust surface deposition processes (kg m^{-2} y^{-1}) in the 1-mode configuration. Wet deposition (**a**, **d**, **g**), dry deposition (**b**, **e**, **h**), and sedimentation (**c**, **f**, **i**) are shown for the control (**a**–**c**) and vegetation-impact (**d**–**f**) simulations. The absolute difference (**g**–**i**) is calculated as control − vegetation-impact.

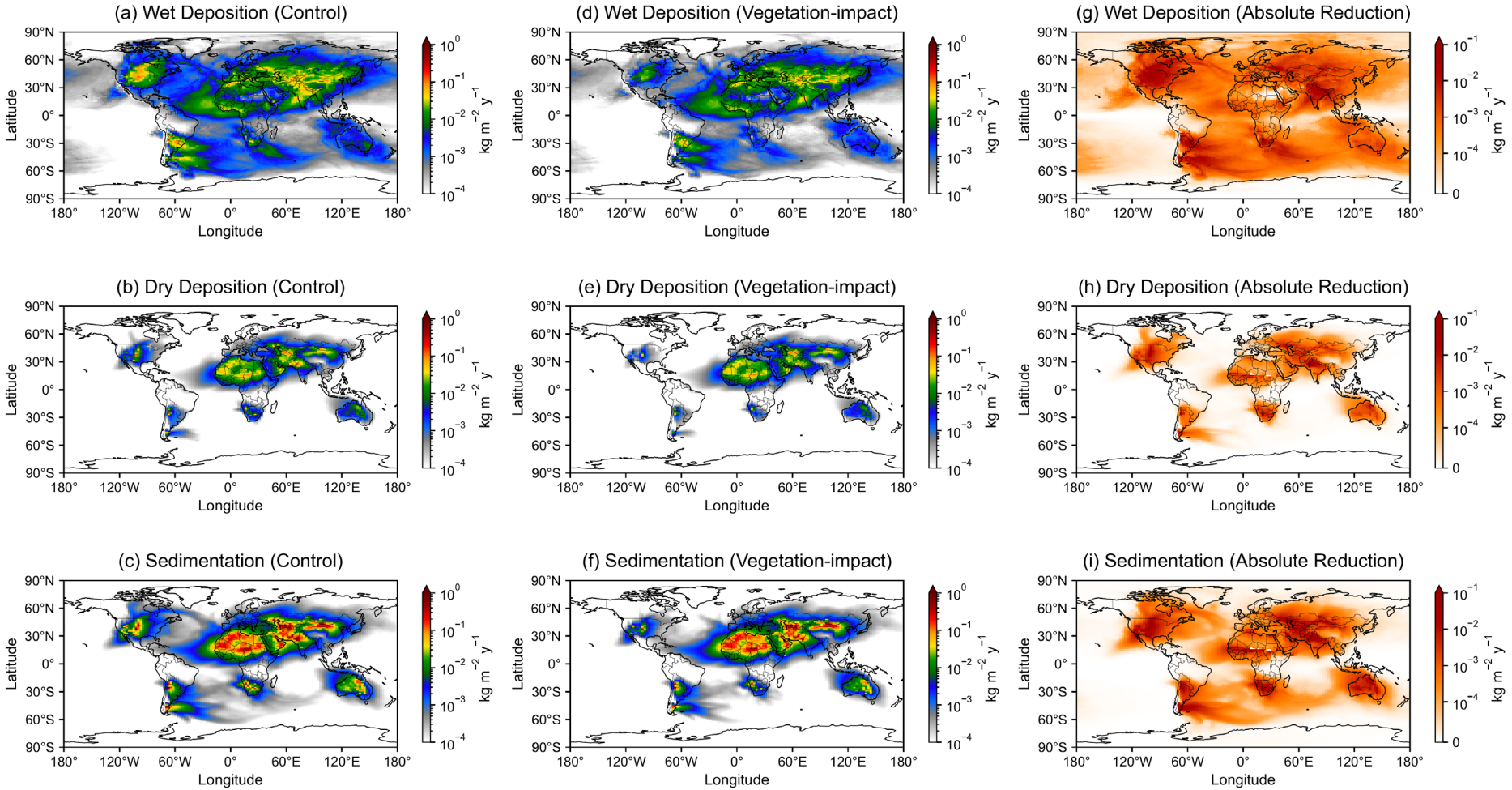

Figure S21. Global dust surface deposition processes (kg m^{-2} y^{-1}) in the 4-mode configuration. Wet deposition (**a**, **d**, **g**), dry deposition (**b**, **e**, **h**), and sedimentation (**c**, **f**, **i**) are shown for the control (**a**–**c**) and vegetation-impact (**d**–**f**) simulations. The absolute difference (**g**–**i**) is calculated as control − vegetation-impact.

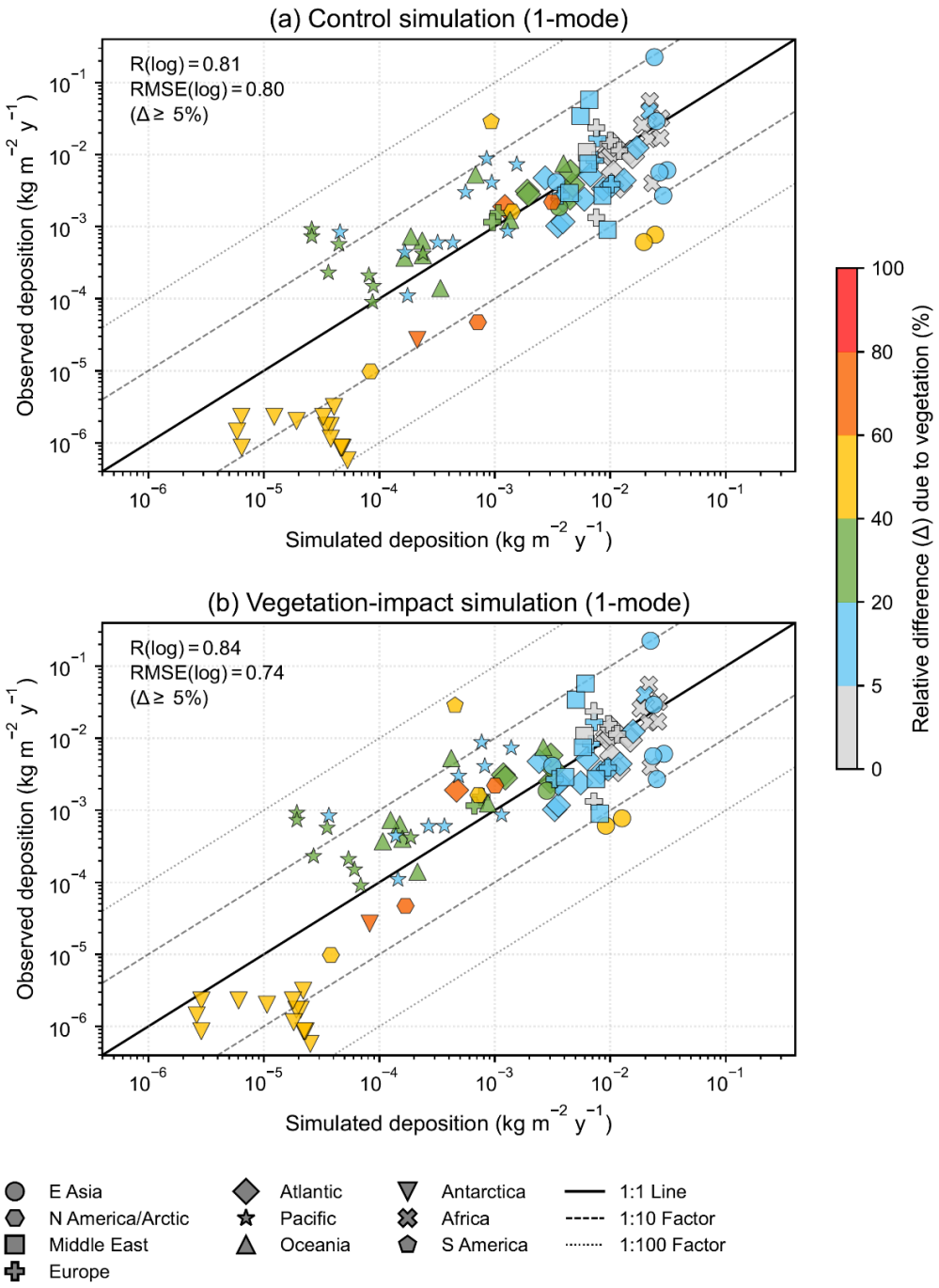

Figure S22. Evaluation of simulated and observed dust surface PM_{10} total deposition (kg m^{-2} y^{-1}) by Albani et al. (2014) in the 1-mode configuration. Panels (**a**) and (**b**) show the control and vegetation-impact simulations, respectively. Both axes are plotted on logarithmic scales. Different regions are indicated by different marker shapes, with colours representing the relative difference between the control and vegetation-impact simulations. Statistical metrics (R and RMSE, computed in log space) are shown in each panel and are calculated only for data points with relative differences (Δ) greater than 5 %, highlighting regions sensitive to vegetation impacts. The solid black line denotes the 1:1 relationship. The inner dashed lines indicate the 10:1 and 1:10 ratios, representing simulated values within one order of magnitude of the observations, while the outer dashed lines represent two orders of magnitude (100:1 and 1:100 ratios). Some observational sites are not visible because their values fall below the lower axis limits.

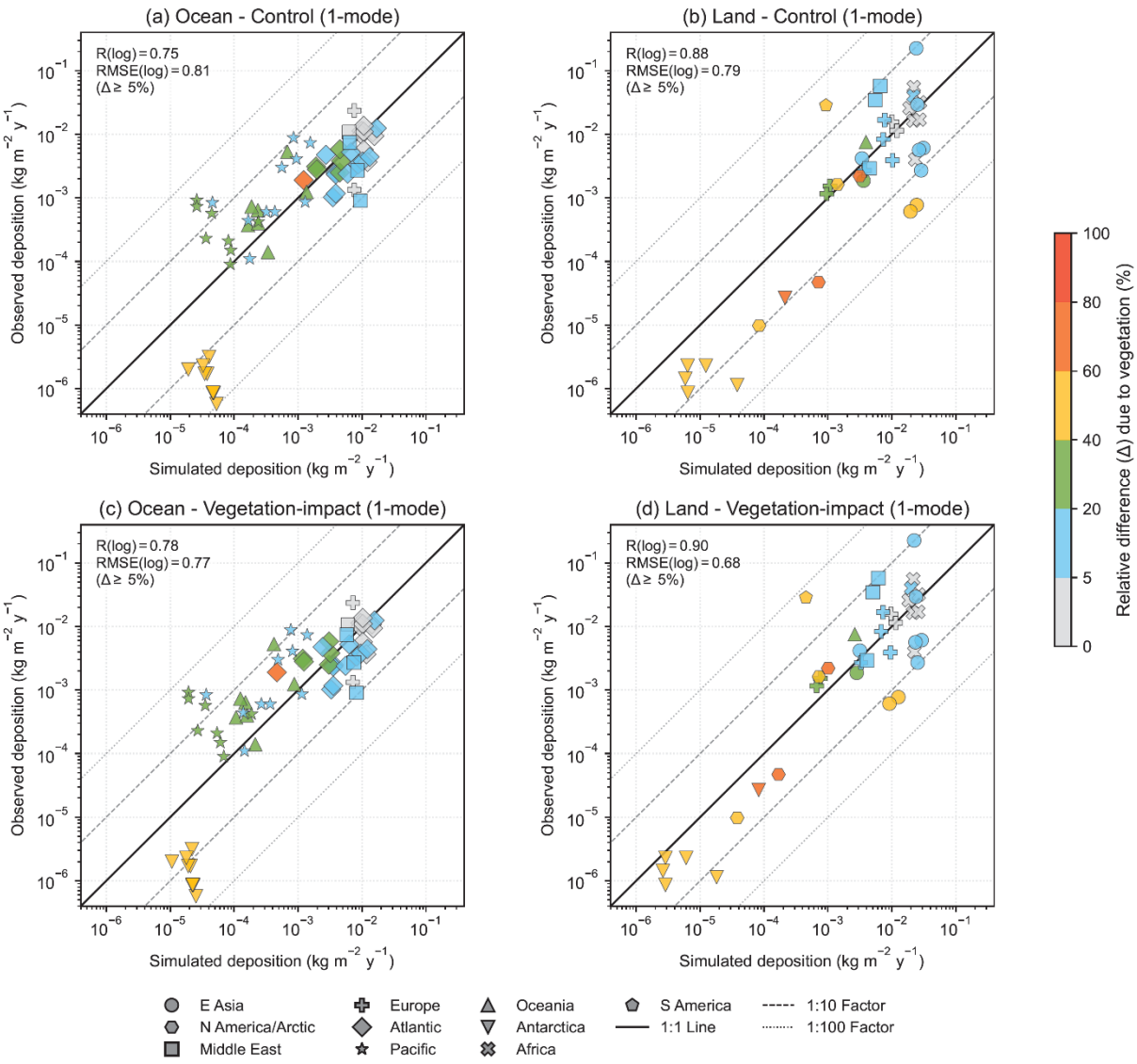

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Figure S23. Comparison of simulated dust total deposition (kg m^{-2} y^{-1}) in the 1-mode configuration with observed dust deposition from Albani et al. (2014), separated into ocean and land sites. Panels (**a**) and (**c**) show ocean points, while panels (**b**) and (**d**) show land points, classified using the NASA land mask. The top row (**a, b**) corresponds to control simulations without vegetation, and the bottom row (**c, d**) to vegetation-impact simulations. Both axes
290 are shown on logarithmic scales. Different regions are indicated by different marker shapes, with colours representing the relative difference between the control and vegetation-impact simulations. Statistical metrics (R and RMSE, computed in log space) are shown in each panel and are calculated only for data points with relative differences (Δ) greater than 5 %, highlighting regions sensitive to vegetation impacts. The solid black line denotes the 1:1 relationship. The inner dashed lines indicate the 10:1 and 1:10 ratios, representing simulated values within
295 one order of magnitude of the observations, while the outer dashed lines represent two orders of magnitude (100:1 and 1:100 ratios). Some observational sites are not visible because their values fall below the lower axis limits.

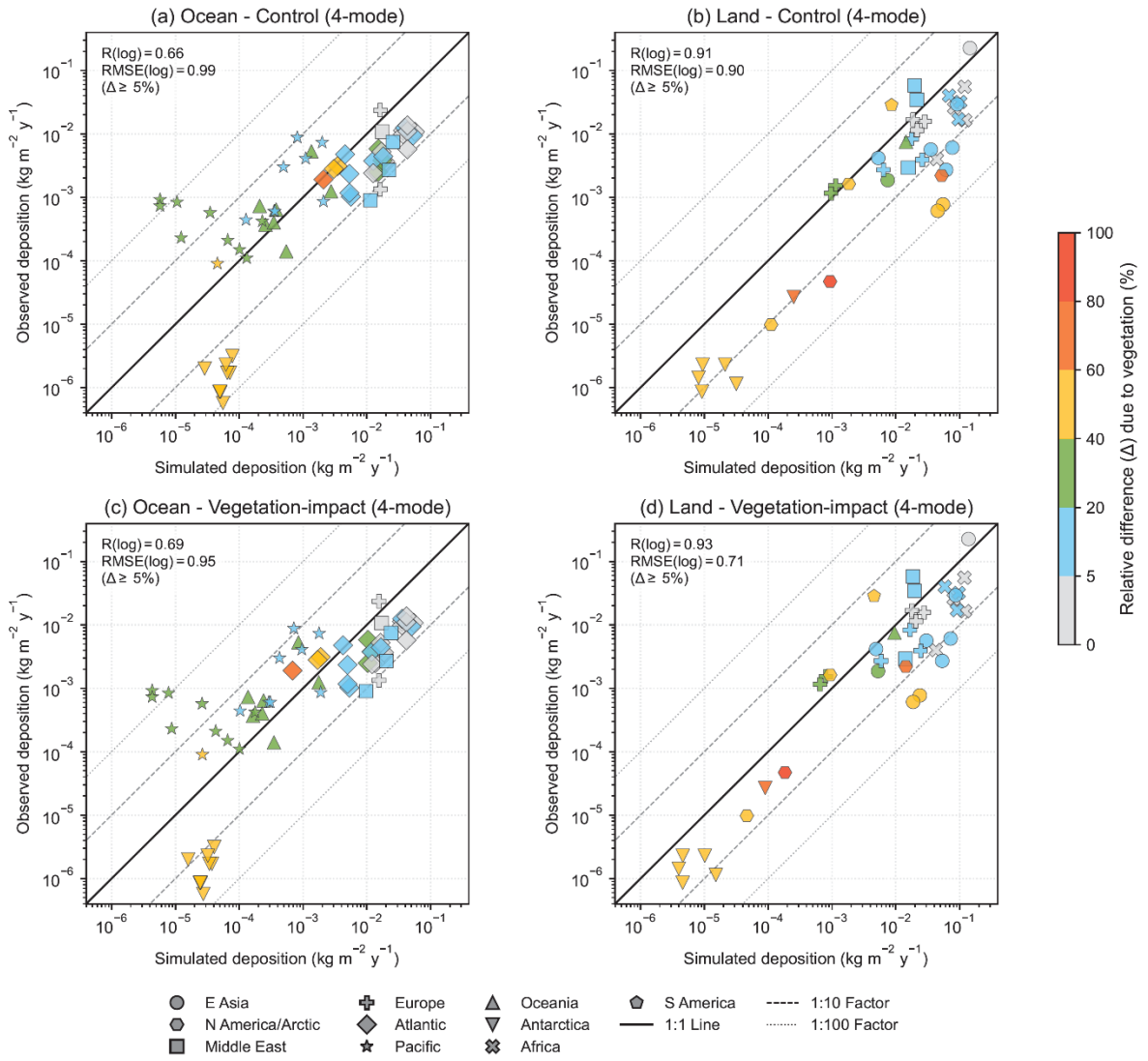

Figure S24. Comparison of simulated dust total deposition (kg m^{-2} y^{-1}) in the 4-mode configuration with observed dust deposition from Albani et al. (2014), separated into ocean and land sites. Panels (**a**) and (**c**) show ocean points, while panels (**b**) and (**d**) show land points, classified using the NASA land mask. The top row (**a, b**) corresponds to control simulations without vegetation, and the bottom row (**c, d**) to vegetation-impact simulations. Both axes are shown on logarithmic scales. Different regions are indicated by different marker shapes, with colours representing the relative difference between the control and vegetation-impact simulations. Statistical metrics (R and RMSE, computed in log space) are shown in each panel and are calculated only for data points with relative differences (Δ) greater than 5 %, highlighting regions sensitive to vegetation impacts. The solid black line denotes the 1:1 relationship. The inner dashed lines indicate the 10:1 and 1:10 ratios, representing simulated values within one order of magnitude of the observations, while the outer dashed lines represent two orders of magnitude (100:1 and 1:100 ratios). Some observational sites are not visible because their values fall below the lower axis limits.

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